

# Dynamics under location uncertainty and other energy-related stochastic subgrid schemes

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# Motivations

- More rigorously identified subgrid dynamics effects
- Quantification of modeling errors (UQ)



Ensemble forecasts and data assimilation

# Contents

- Models under location uncertainty (LU)
- Some parameterization of the models under location uncertainty
- A new energy-budget-based stochastic scheme: WaveHyperv
- Numerical comparisons

# Part I

## Models under location uncertainty (LU)

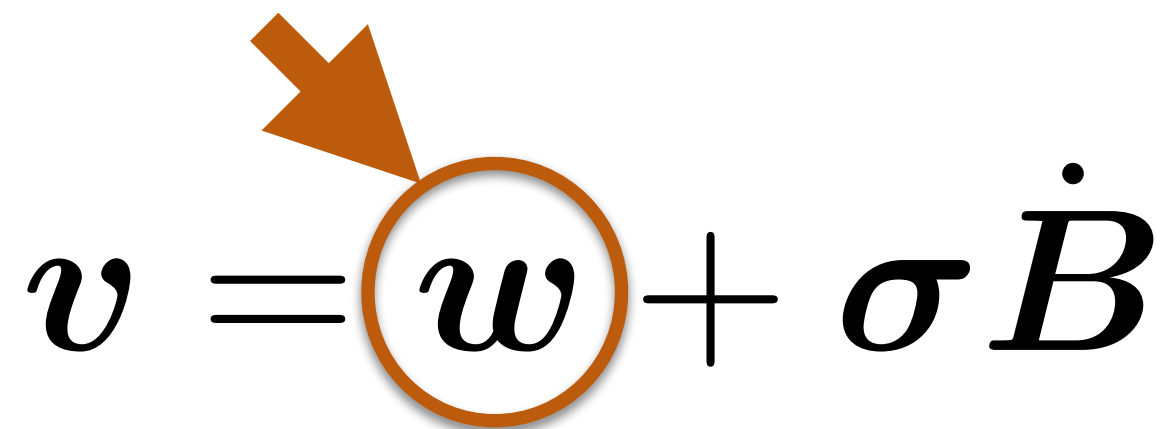


LU : Adding  
random velocity

$$\boldsymbol{v} = \boldsymbol{w} + \sigma \dot{\boldsymbol{B}}$$

# LU : Adding random velocity

Resolved  
large scales



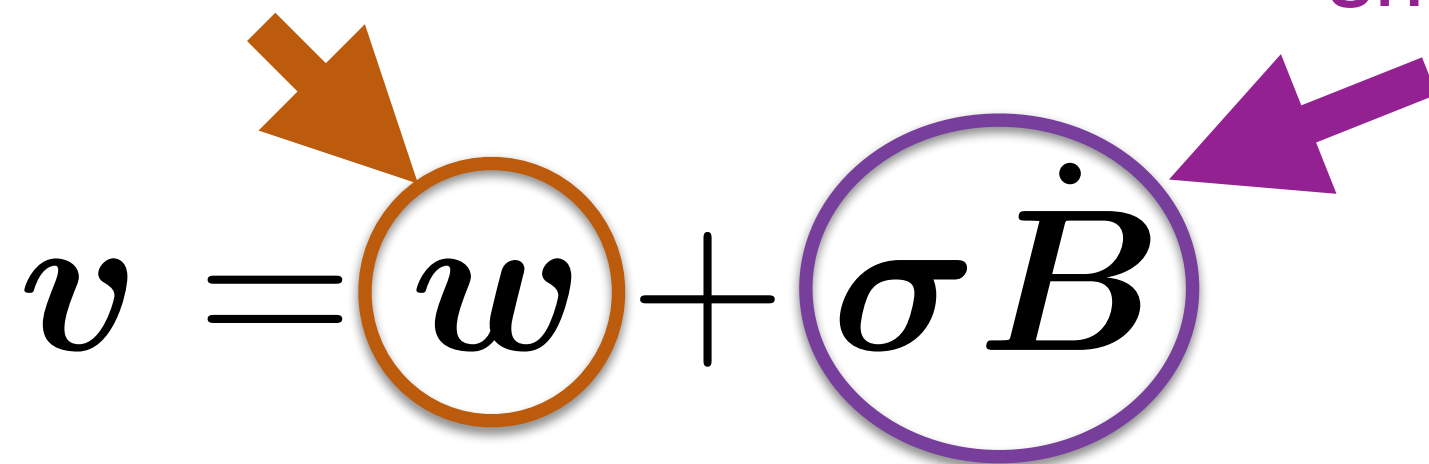
The diagram shows the equation  $v = w + \sigma \dot{B}$ . The variable  $w$  is enclosed in a brown circle, and a brown arrow points from the text "Resolved large scales" to this circle.

$$v = w + \sigma \dot{B}$$

# LU : Adding random velocity

Resolved  
large scales

White-in-time  
small scales



The diagram shows the equation  $v = w + \sigma \dot{B}$ . The term  $w$  is enclosed in an orange circle, and an orange arrow points from the text "Resolved large scales" to it. The term  $\sigma \dot{B}$  is enclosed in a purple circle, and a purple arrow points from the text "White-in-time small scales" to it.

$$v = w + \sigma \dot{B}$$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# LU : Adding random velocity

Resolved  
large scales

White-in-time  
small scales

$$v = w + \sigma \dot{B}$$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# LU : Adding random velocity

Resolved  
large scales

White-in-time  
small scales

$$v = w + \sigma \dot{B}$$

References : Mikulevicius &  
Rozovskii, 2004  
Flandoli, 2011

**Memin, 2014**  
Resseguier et al. 2017 a, b, c  
Cai et al. 2017  
Chapron et al. 2018  
Yang & Memin 2019  
Resseguier et al. 2019 a,b

**Holm, 2015**  
Holm and  
Tyranowski, 2016  
Arnaudon et al. 2017  
Cotter and al. 2017

Crisan et al., 2017  
Gay-Balmaz & Holm 2017  
Cotter and al. 2018 a, b  
Cotter and al. 2019

# LU : Adding random velocity

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

Resolved  
large scales

White-in-time  
small scales

$$v = w + \sigma \dot{B}$$

LU

**Memin, 2014**

Resseguier et al. 2017 a, b, c

Cai et al. 2017

Chapron et al. 2018

Yang & Memin 2019

Resseguier et al. 2019 a,b

SALT

**Holm, 2015**

Holm and

Tyranowski, 2016

Arnaudon et al. 2017

Cotter and al. 2017

Crisan et al., 2017

Gay-Balmaz & Holm 2017

Cotter and al. 2018 a, b

Cotter and al. 2019

References : Mikulevicius &  
Rozovskii, 2004  
Flandoli, 2011

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

$$\frac{D\Theta}{Dt} = 0$$

Large scales:

$$\boldsymbol{w}$$

Small scales:

$$\sigma \dot{\boldsymbol{B}}$$

Variance  
tensor:

$$\boldsymbol{a} = \boldsymbol{a}(\boldsymbol{x}, \boldsymbol{x}) = \frac{\mathbb{E}\{\sigma d\boldsymbol{B} (\sigma d\boldsymbol{B})^T\}}{dt}$$

# Advection of tracer $\Theta$



Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

$$\partial_t \Theta + w^\star \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta = \nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)$$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

$$\partial_t \Theta + \boxed{\text{Advection}} \quad w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta = \nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)$$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
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# Advection of tracer $\Theta$

$$\partial_t \Theta + \underbrace{w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta}_{\text{Advection}} = \underbrace{\nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)}_{\text{Diffusion}}$$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

$$\partial_t \Theta + \underbrace{w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta}_{\text{Advection}} = \underbrace{\nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)}_{\text{Diffusion}}$$

Drift correction

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

Multiplicative  
random  
forcing

$$\partial_t \Theta + \underbrace{w^\star \cdot \nabla \Theta}_{\text{Advection}} + \sigma \dot{B} \cdot \nabla \Theta = \underbrace{\nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)}_{\text{Diffusion}}$$

Drift correction

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

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Drift correction

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Advection of tracer $\Theta$

Multiplicative  
random  
forcing

$$\partial_t \Theta + w^* \cdot \nabla \Theta + \sigma \dot{B} \cdot \nabla \Theta = \nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)$$

Drift correction

Advection

Diffusion

# Advection of tracer $\Theta$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

Multiplicative  
random  
forcing

Balanced  
energy  
exchanges

$$\partial_t \Theta + \underbrace{w^\star \cdot \nabla \Theta}_{\text{Advection}} + \underbrace{\sigma \dot{B} \cdot \nabla \Theta}_{\text{Diffusion}} = \nabla \cdot \left( \frac{1}{2} a \nabla \Theta \right)$$

Drift correction



# Part II

## Some parameterization of LU models

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## Some parameterization of LU models

$\sigma = ?$

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

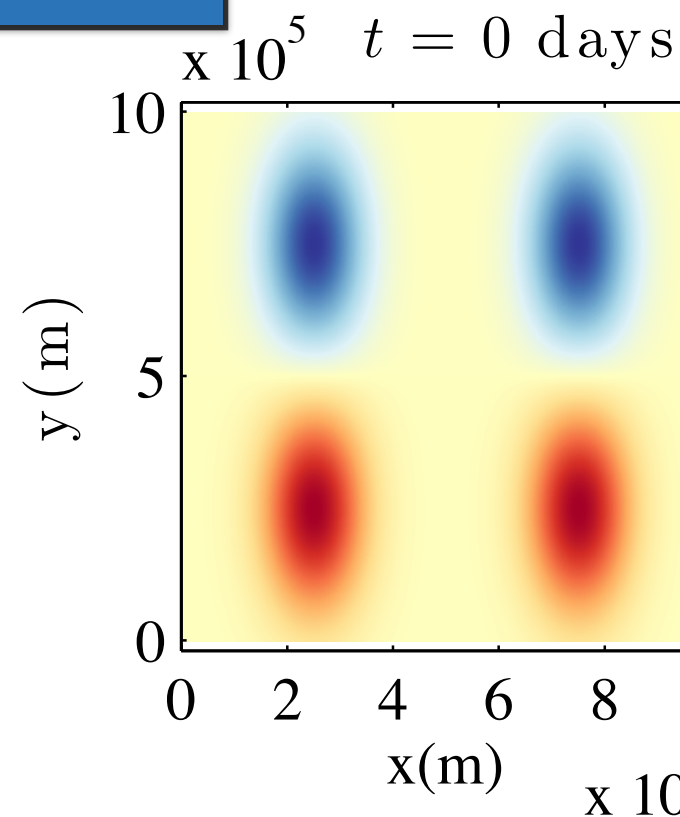
Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

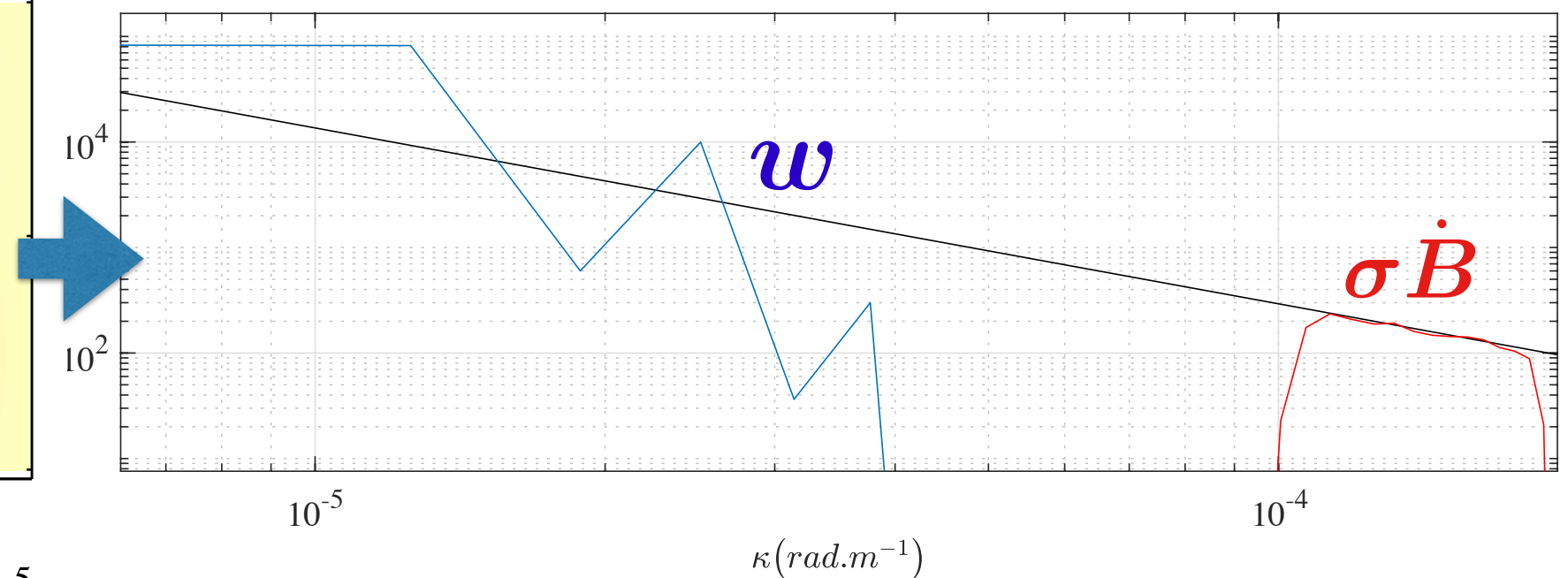
MU Spec

# Spectral model

(homogeneous and stationary  $\sigma \dot{B}$ )



## KE Spectrum



Code online

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

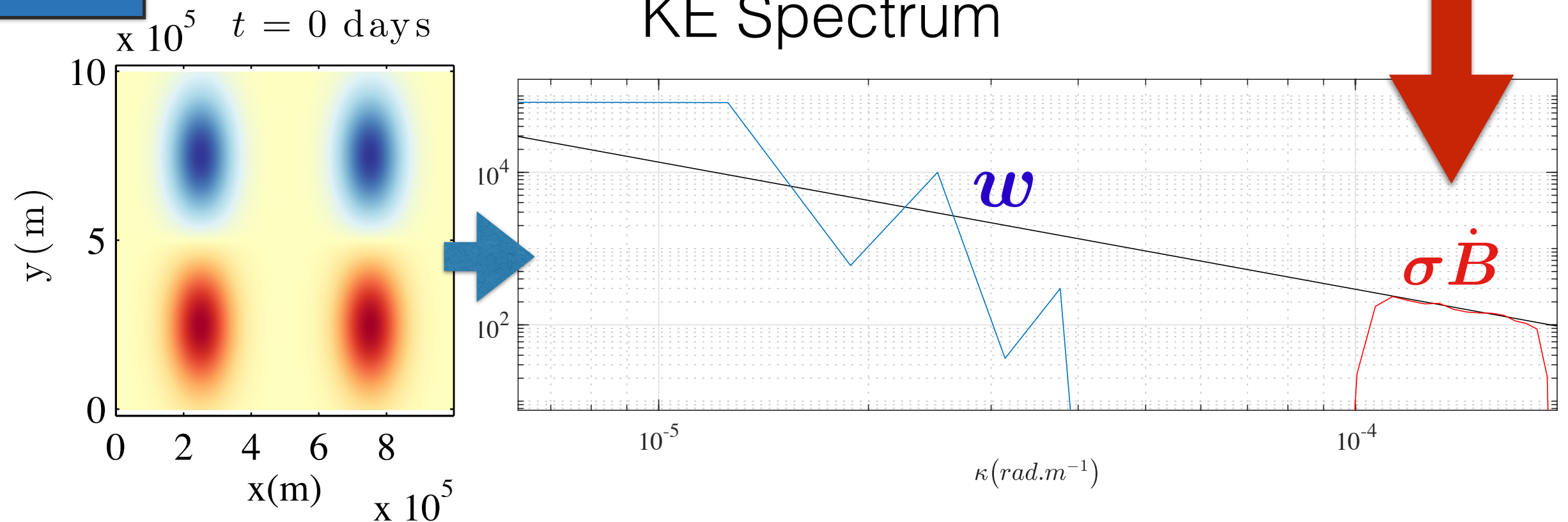
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

# Spectral model

(homogeneous and stationary  $\sigma \dot{B}$ )

MU Spec

Fixed spectrum  
at small scales



Code online

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

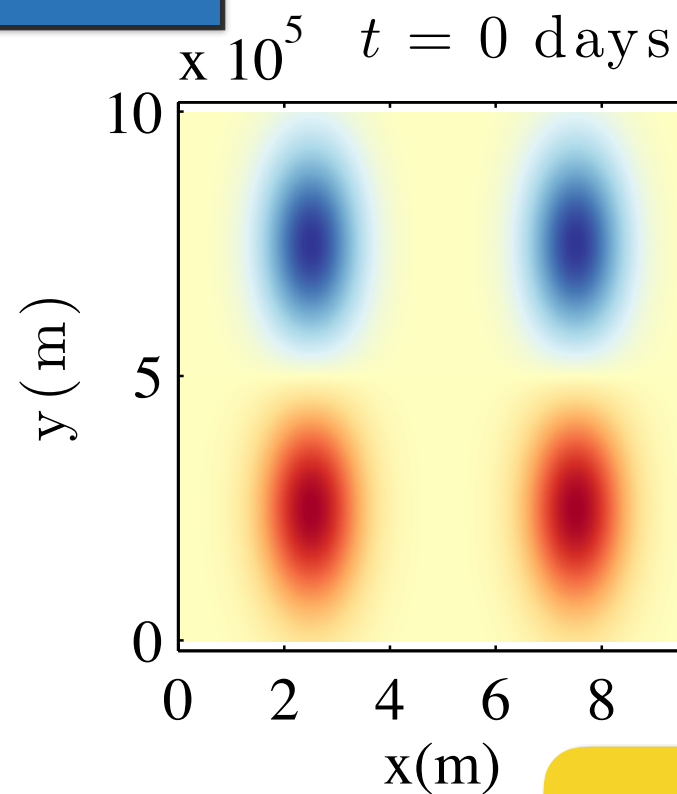
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

MU Spec

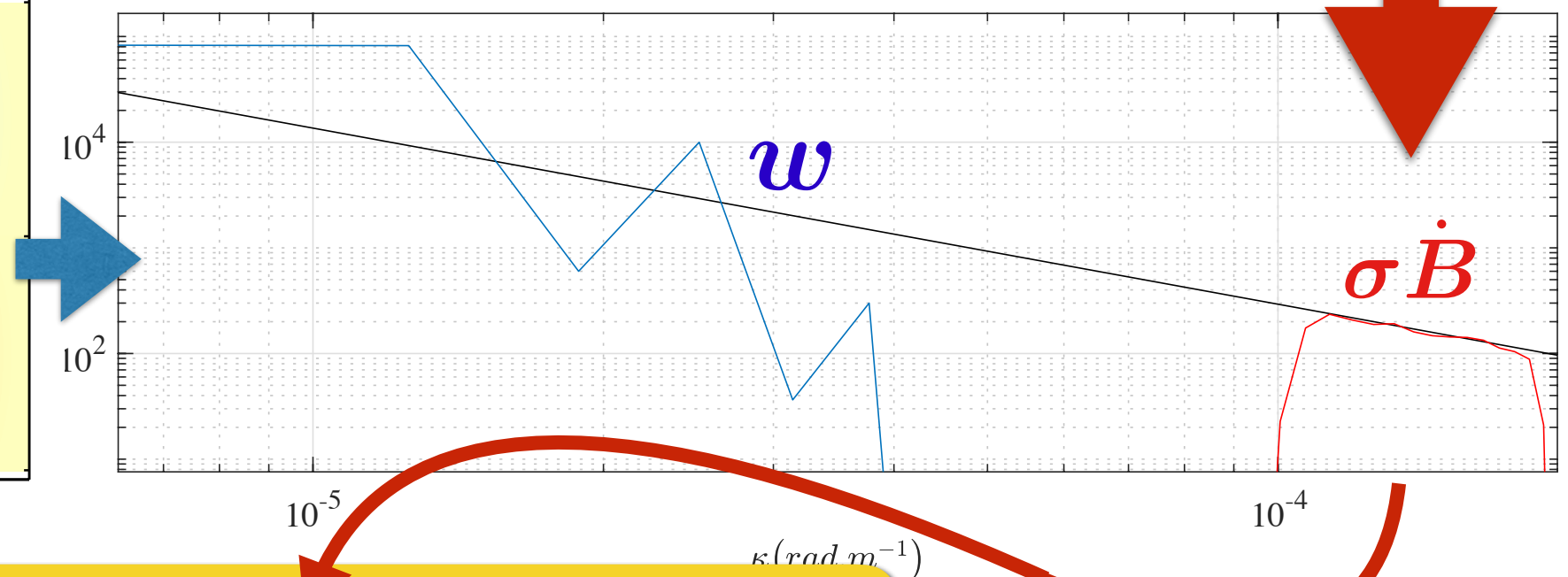
# Spectral model

(homogeneous and stationary  $\sigma \dot{B}$ )

Fixed spectrum  
at small scales



## KE Spectrum



$$\sigma \dot{B} = (\text{filter}) * (\text{white noise})$$

Code online

Reference:  
*Resseguier, Memin & Chapron 2017b*

# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

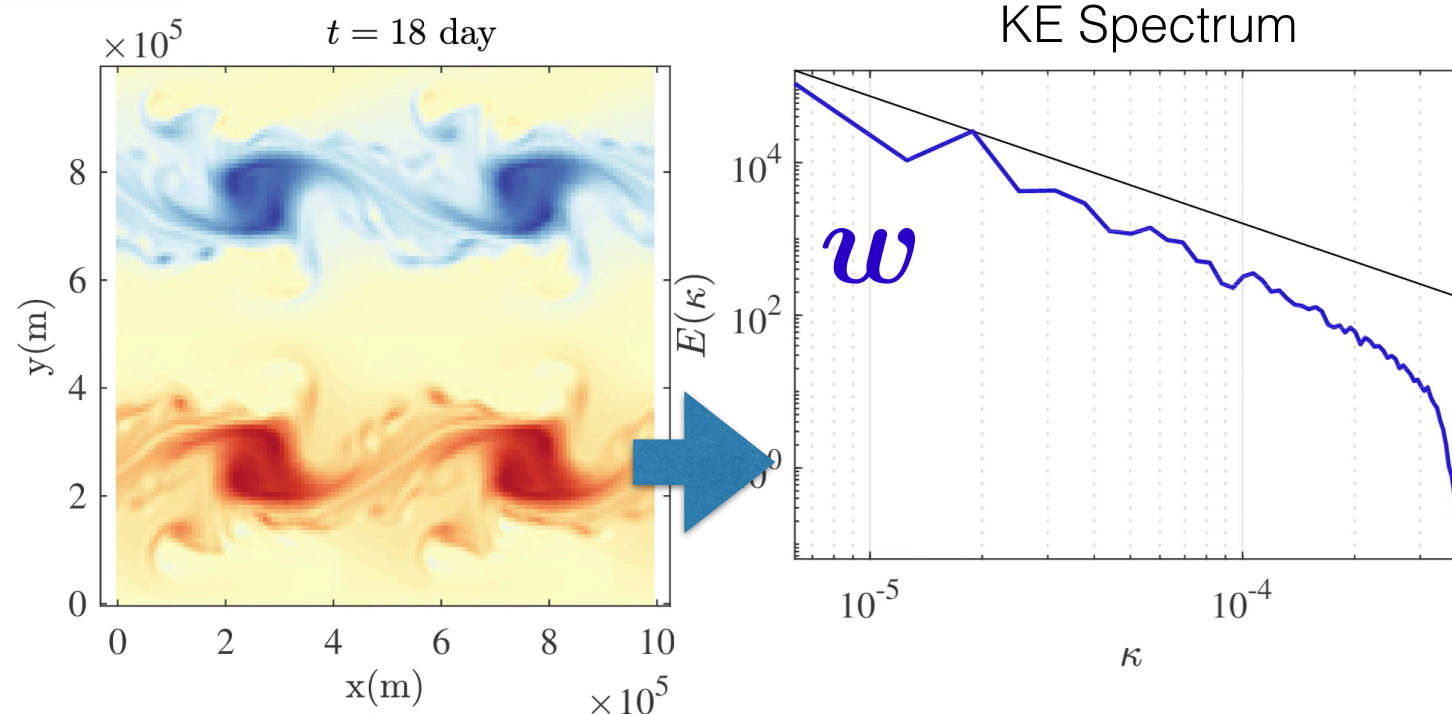
Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )



# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

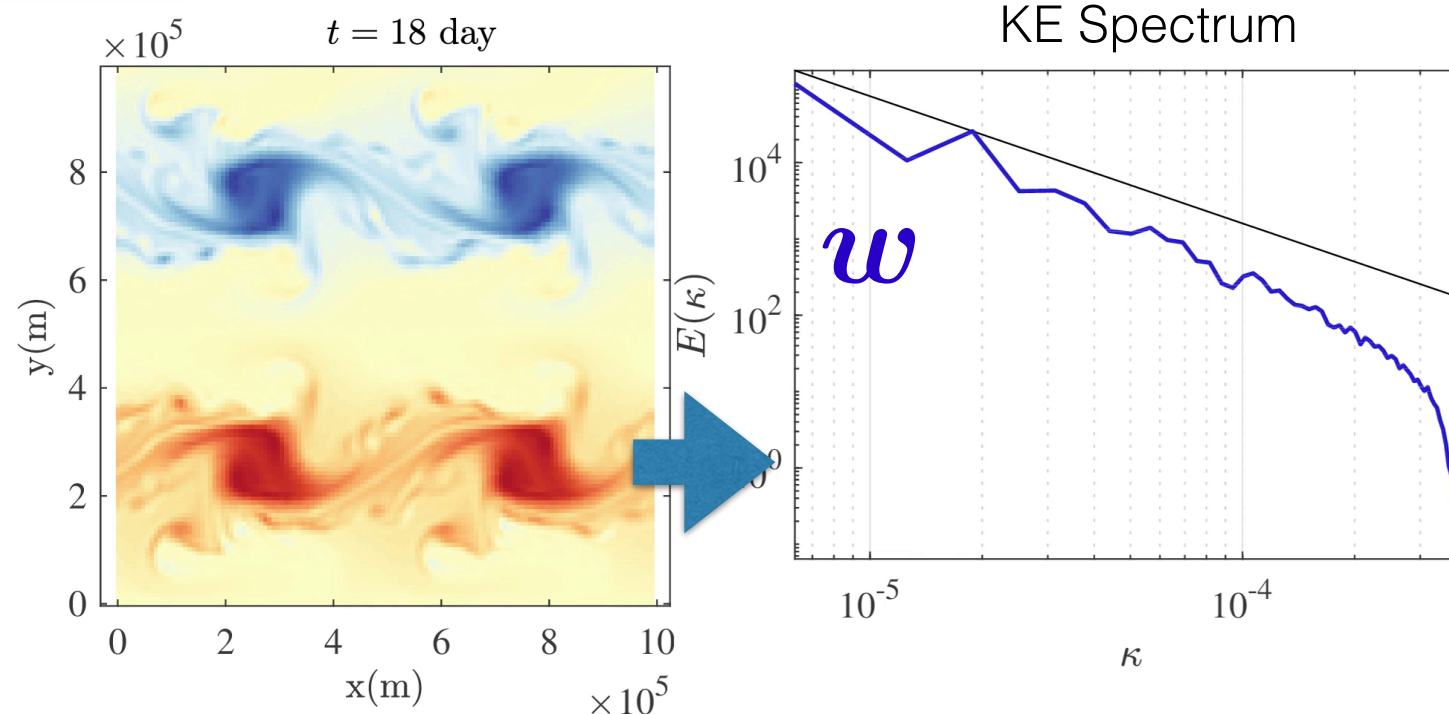
Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )

Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa)\tau(\kappa)$$



# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

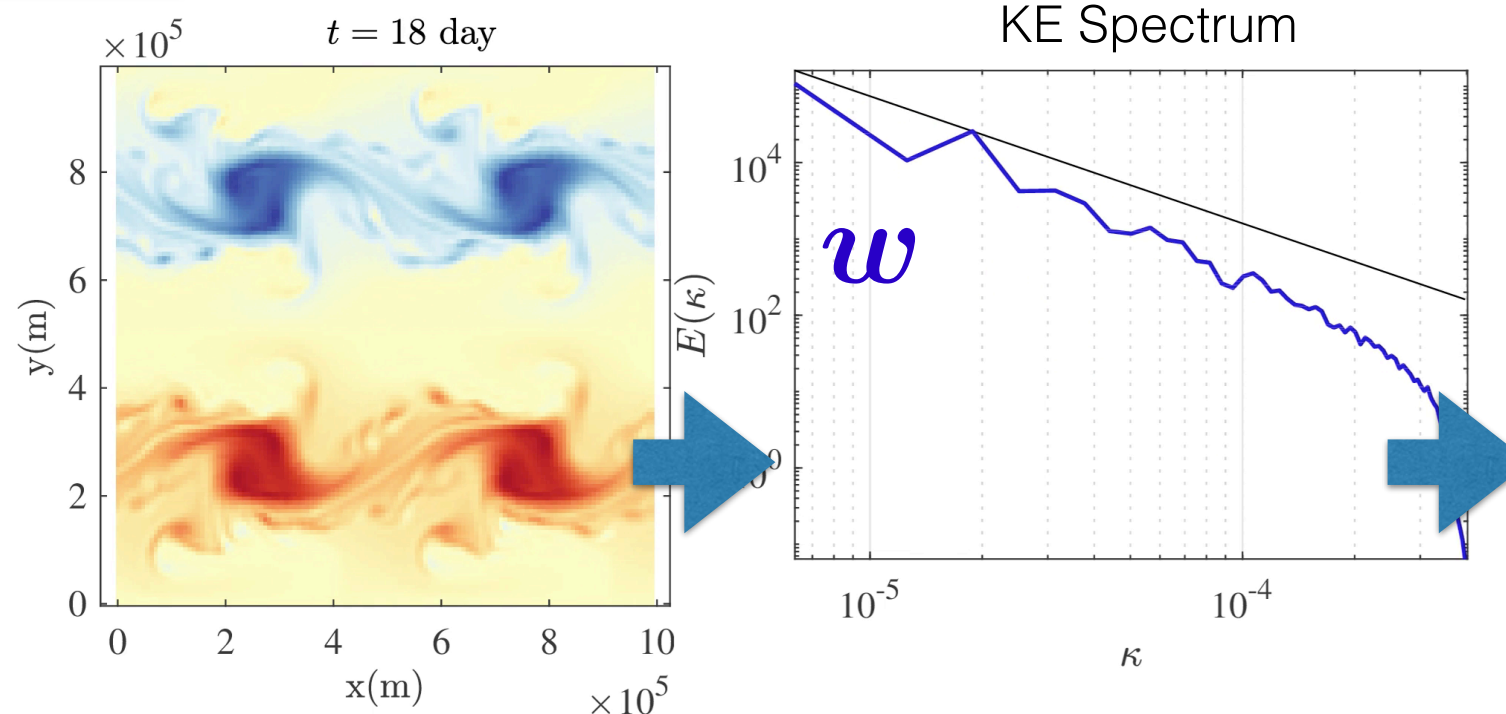
Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )

Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa)\tau(\kappa)$$





# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

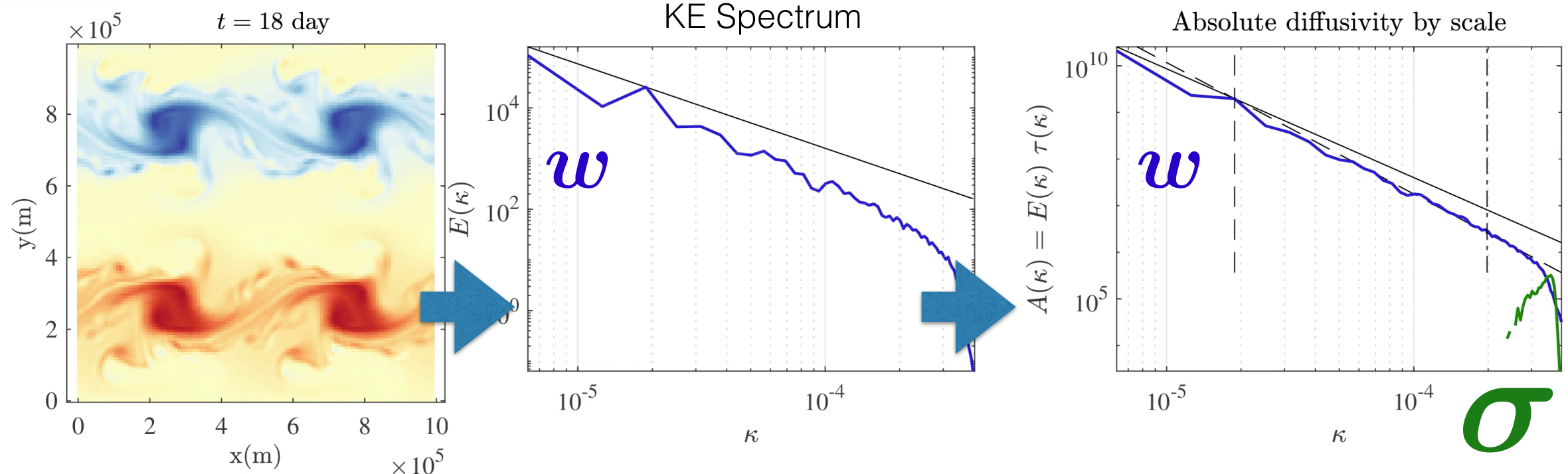
Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )

Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa) \tau(\kappa)$$



# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

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Small scales:

$\sigma \dot{B}$

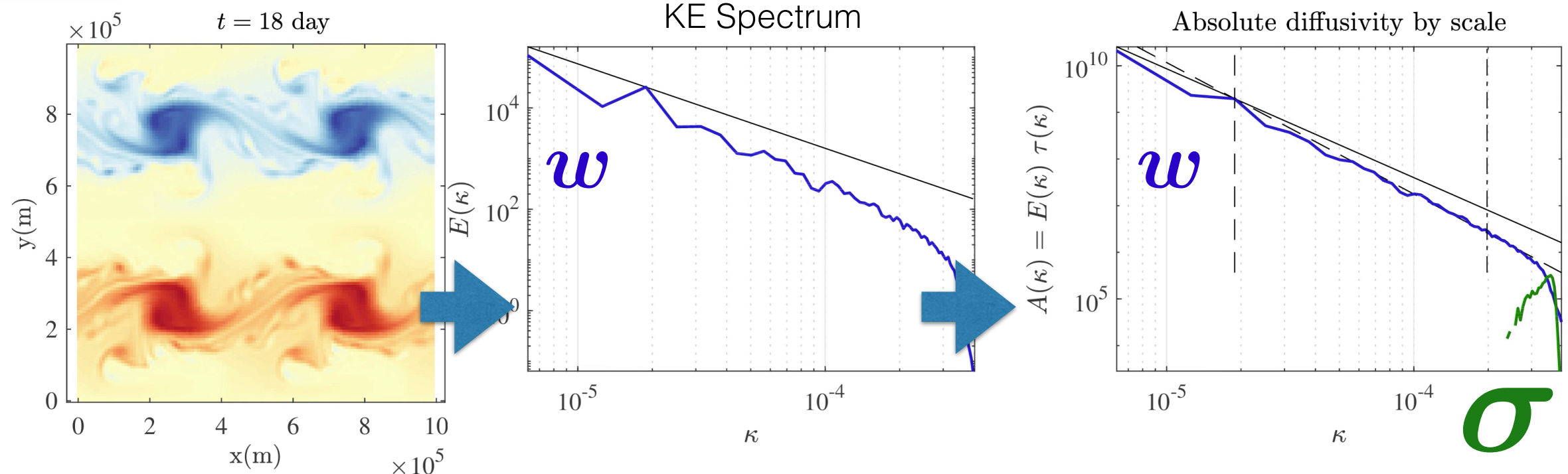
Variance  
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Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa) \tau(\kappa)$$



# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

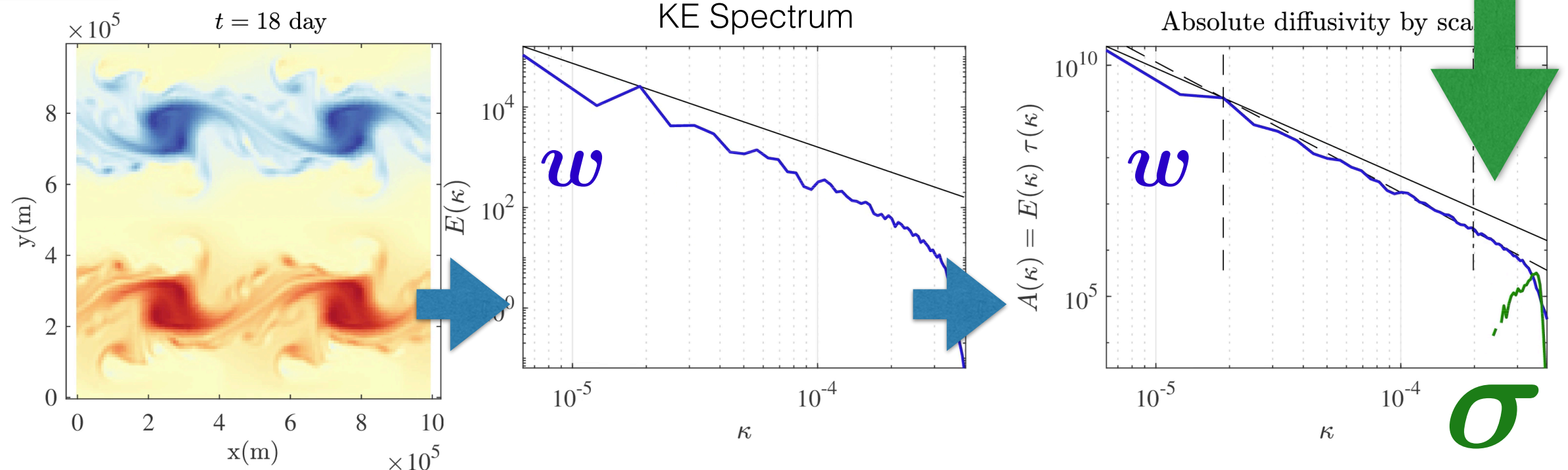
$$\mathbf{a} = \mathbf{a}(\mathbf{x}, \mathbf{x}) = \frac{\mathbb{E}\{\sigma d\mathbf{B} (\sigma d\mathbf{B})^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )

Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa) \tau(\kappa)$$

Residual  
non-stationary  
ADSD



# Absolute Diffusivity Spectral Density

MU ADSD

Large scales:

$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

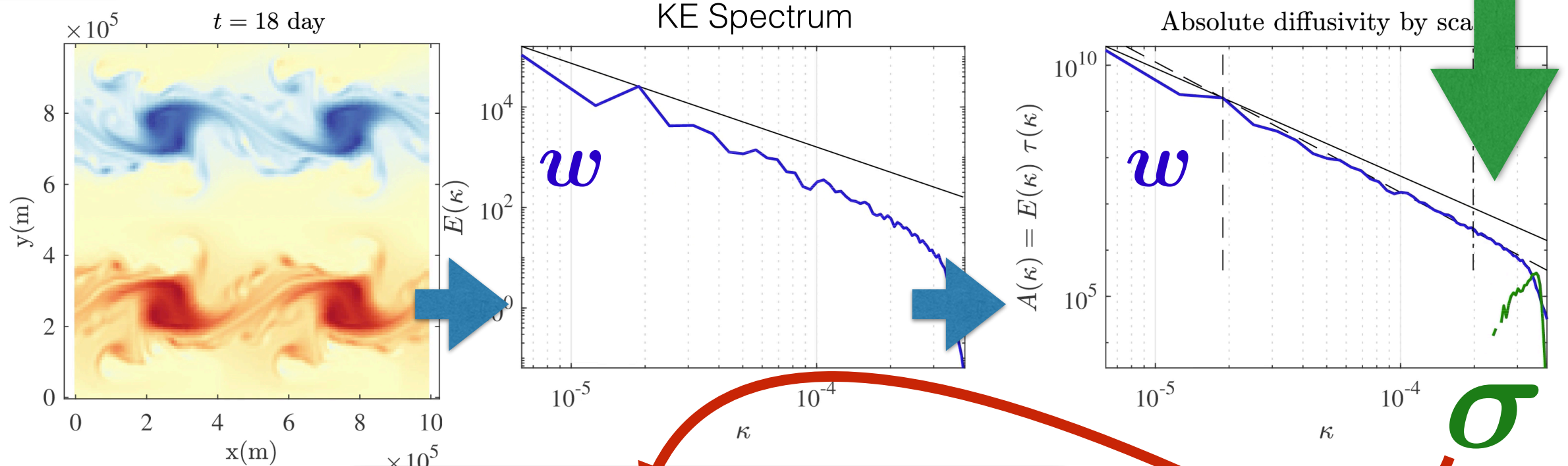
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$

(homogeneous but non-stationary and tuning-free  $\sigma \dot{B}$ )

Absolute Diffusivity  
Spectral Density

$$A(\kappa) = E(\kappa) \tau(\kappa)$$

Residual  
non-stationary  
ADSD



$$\sigma \dot{B} = (\text{filter}) * (\text{white noise})$$

# Random switching of points

MU SVD

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

Large scales:

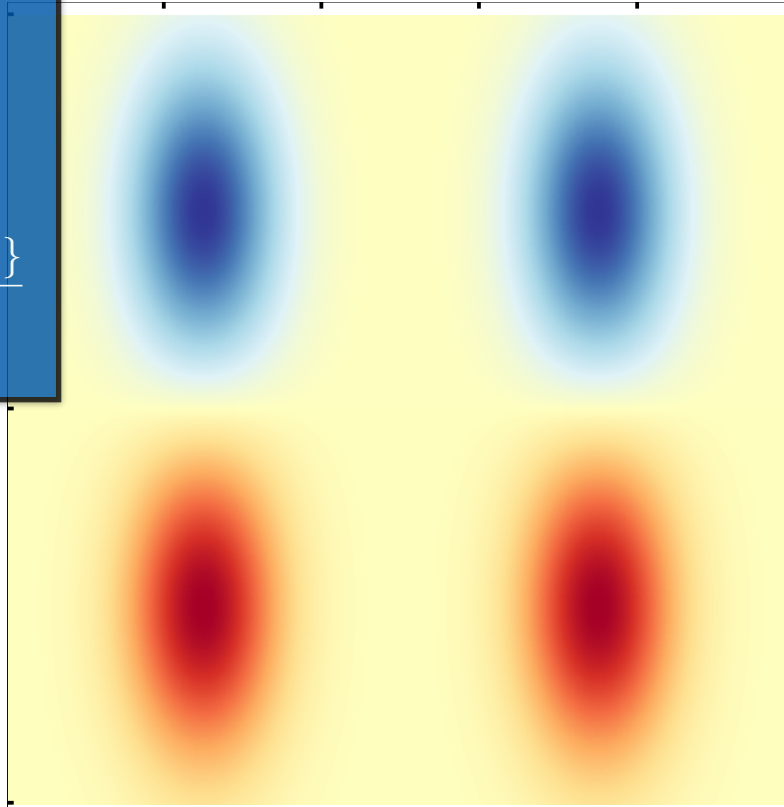
$w$

Small scales:

$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$





# Random switching of points

MU SVD

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

Large scales:

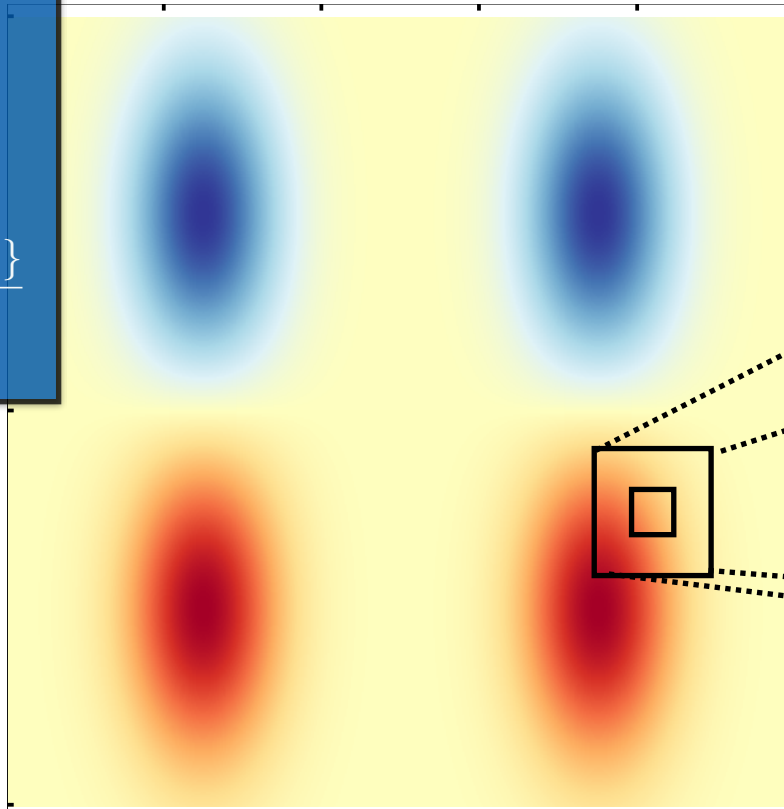
$w$

Small scales:

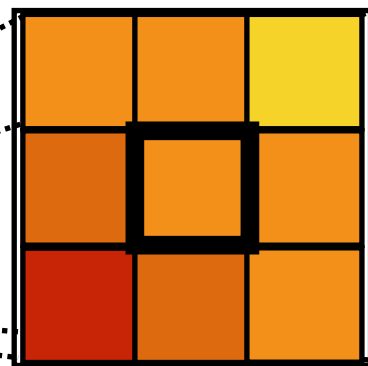
$\sigma \dot{B}$

Variance  
tensor:

$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



Neighbor



# Random switching of points

MU SVD

Large scales:

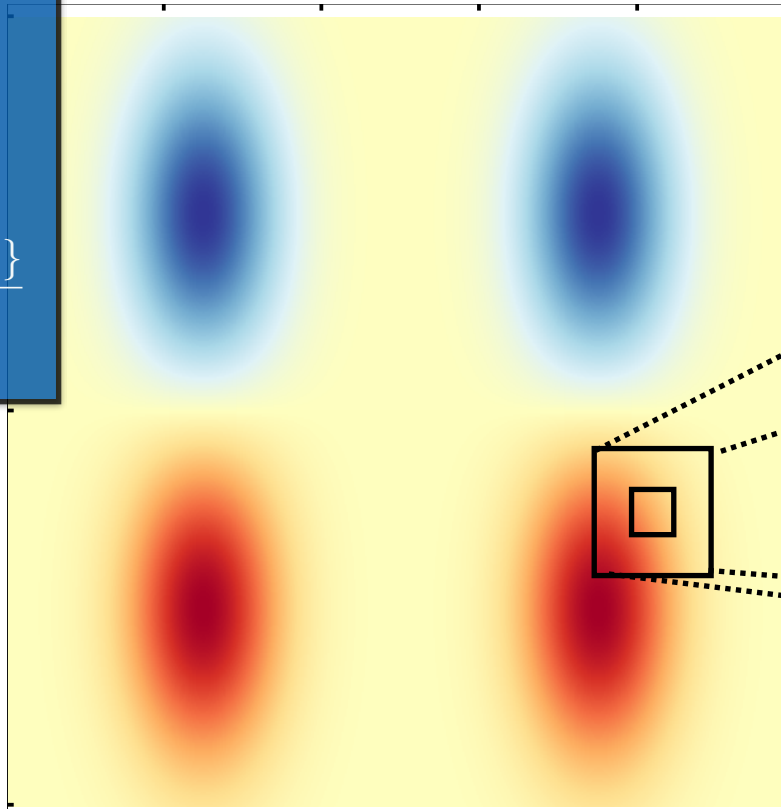
$w$

Small scales:

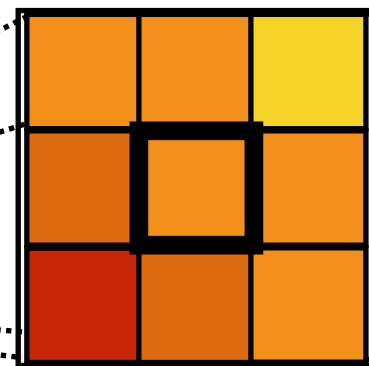
$\sigma \dot{B}$

Variance  
tensor:

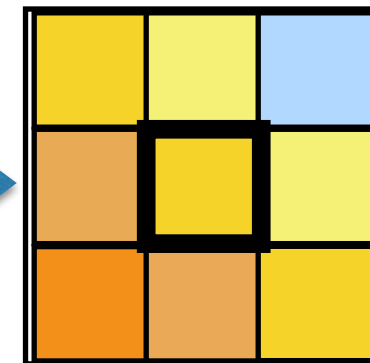
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



Neighbor



Centered  
neighbor



# Random switching of points

MU SVD

Large scales:

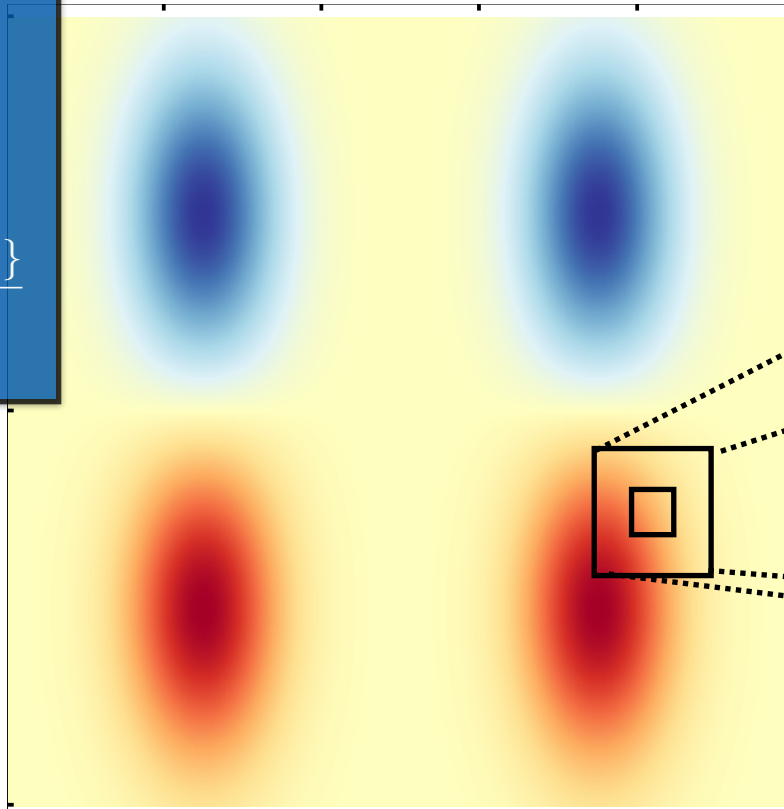
$w$

Small scales:

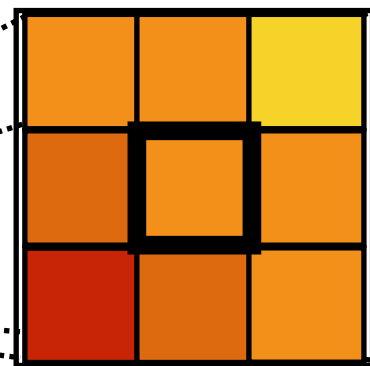
$\sigma \dot{B}$

Variance  
tensor:

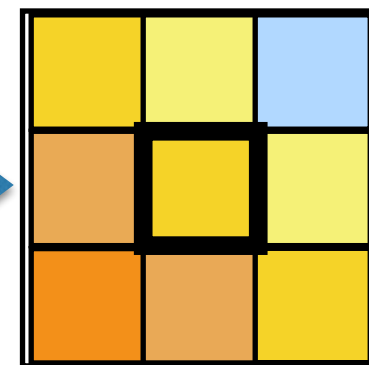
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



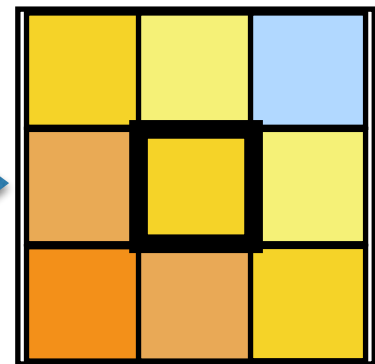
Neighbor



Centered  
neighbor



Random  
selection





# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

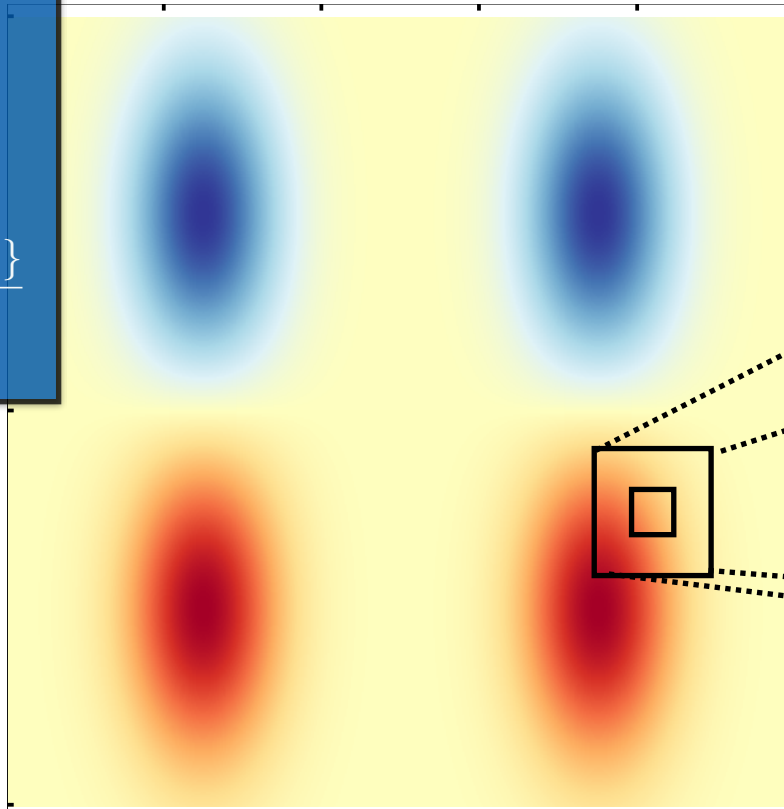
$w$

Small scales:

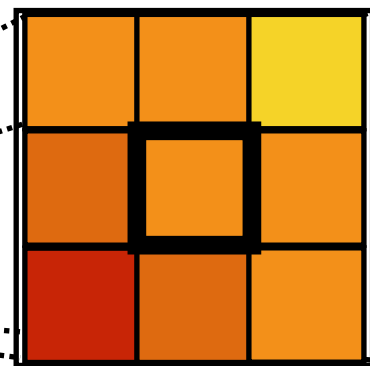
$\sigma \dot{B}$

Variance  
tensor:

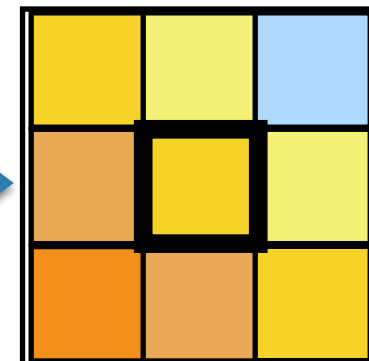
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



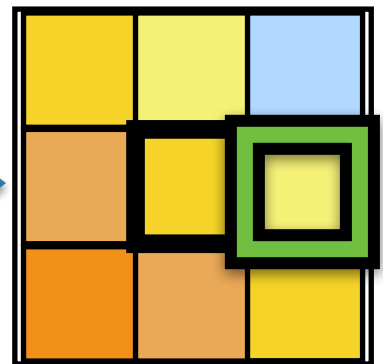
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

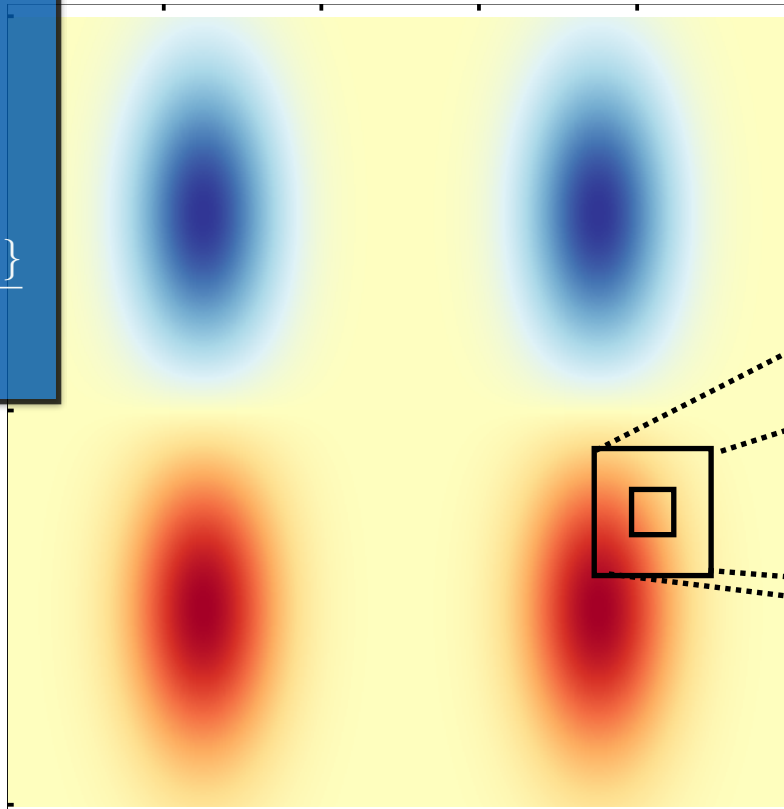
$w$

Small scales:

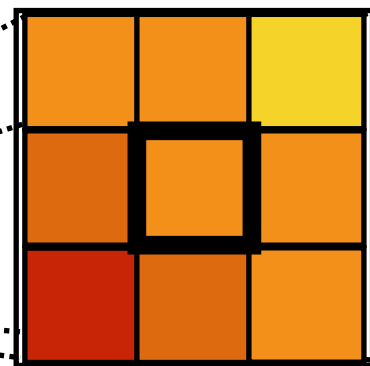
$\sigma \dot{B}$

Variance  
tensor:

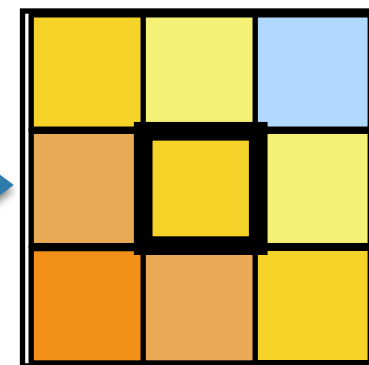
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



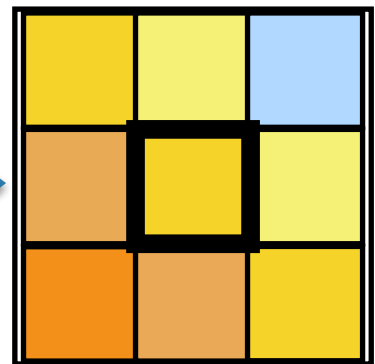
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

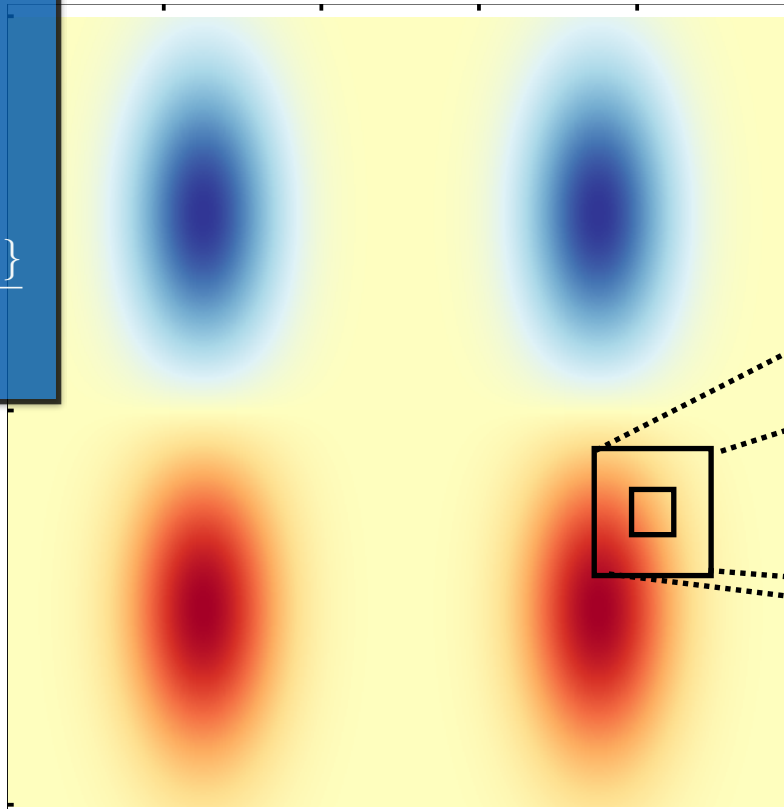
$w$

Small scales:

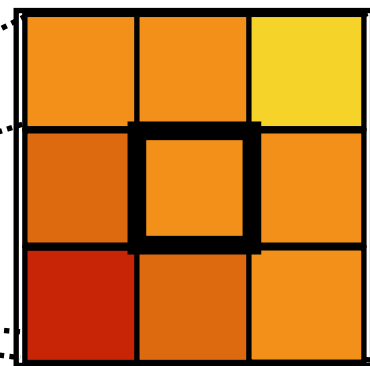
$\sigma \dot{B}$

Variance  
tensor:

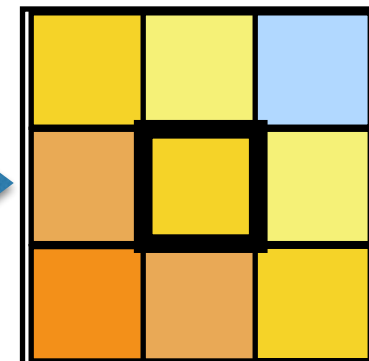
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



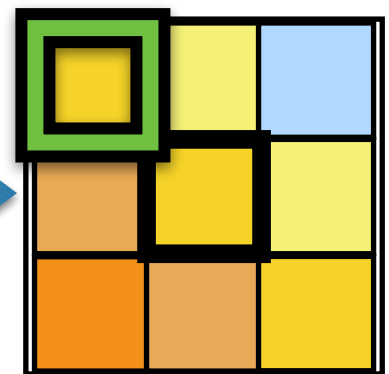
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

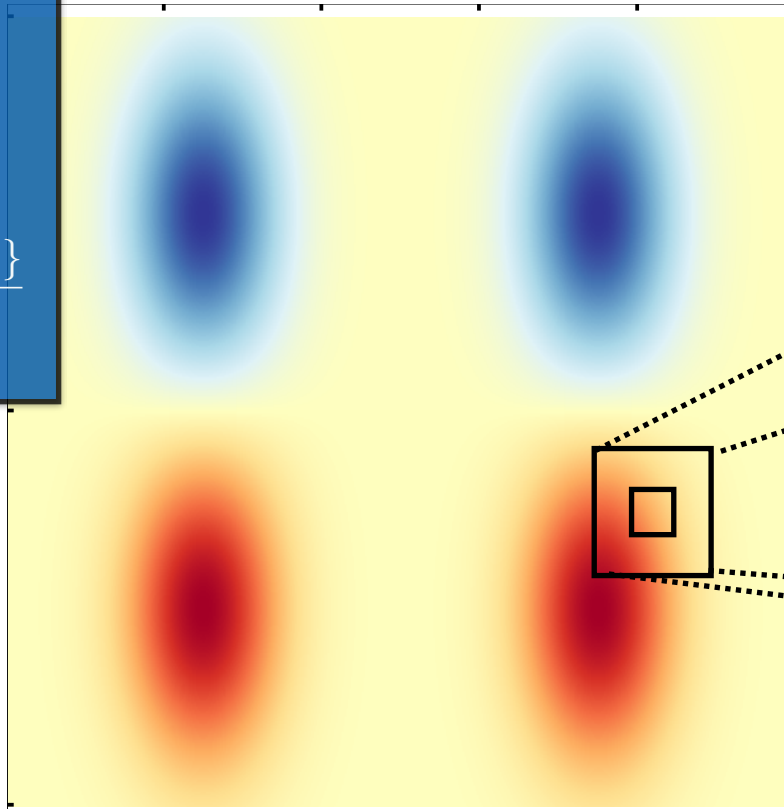
$w$

Small scales:

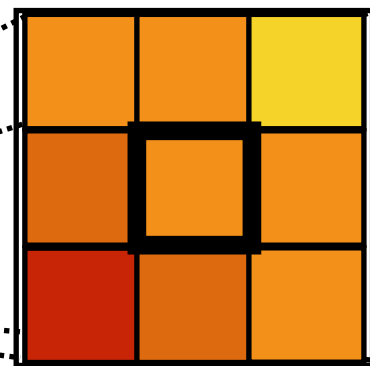
$\sigma \dot{B}$

Variance  
tensor:

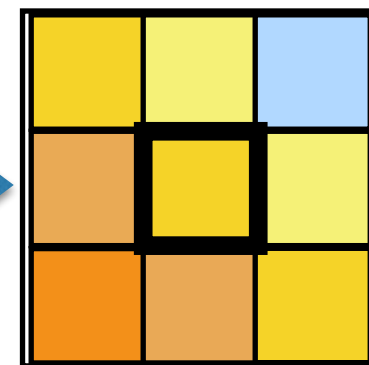
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



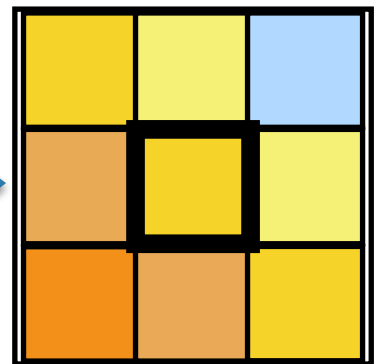
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

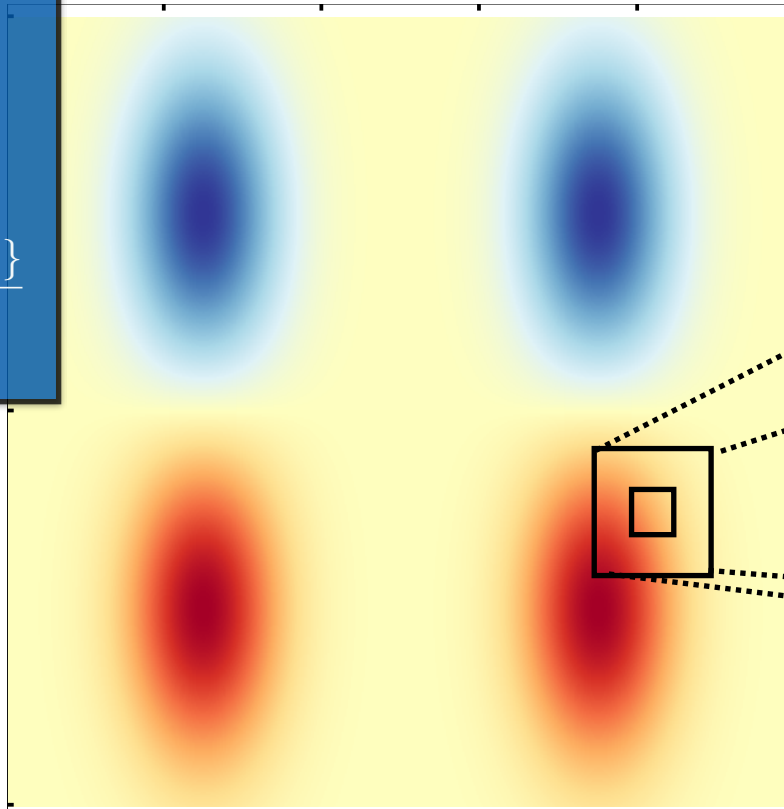
$w$

Small scales:

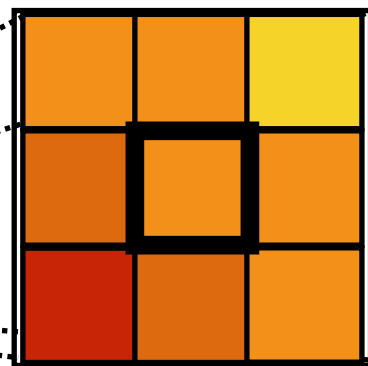
$\sigma \dot{B}$

Variance  
tensor:

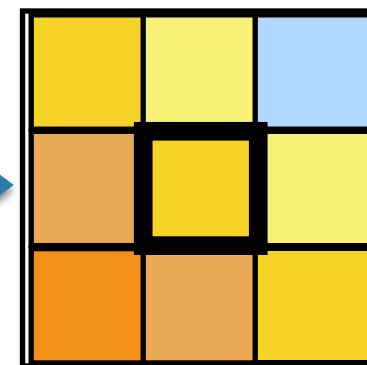
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



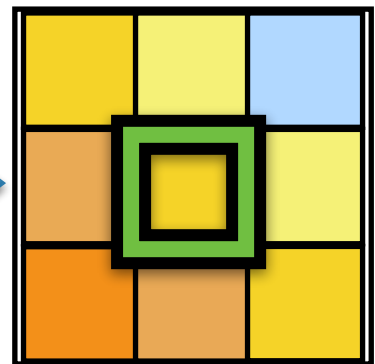
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

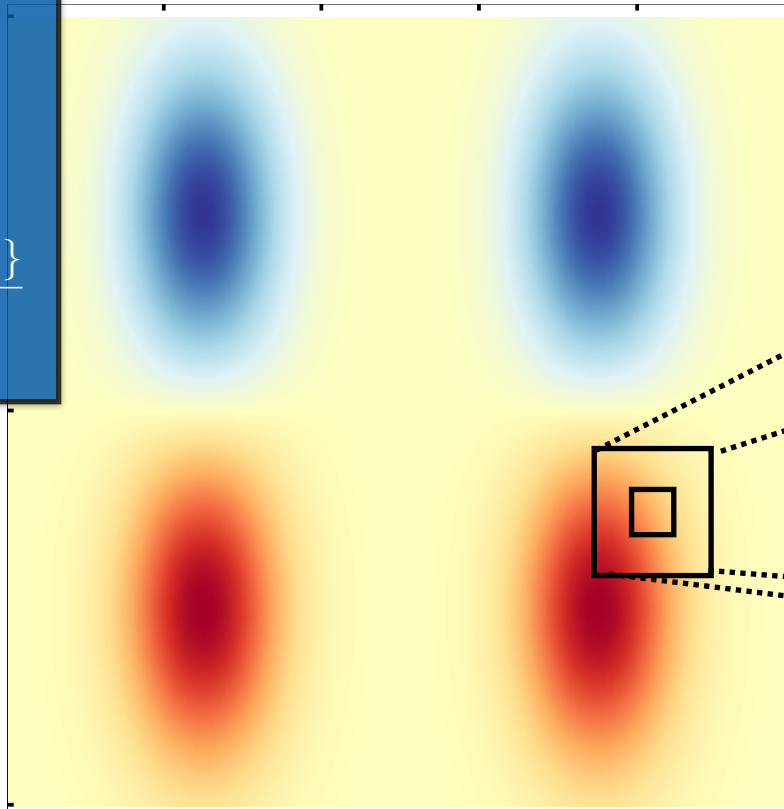
$w$

Small scales:

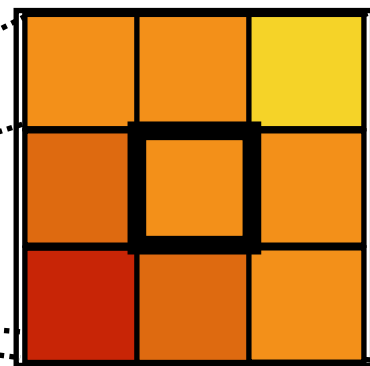
$\sigma \dot{B}$

Variance  
tensor:

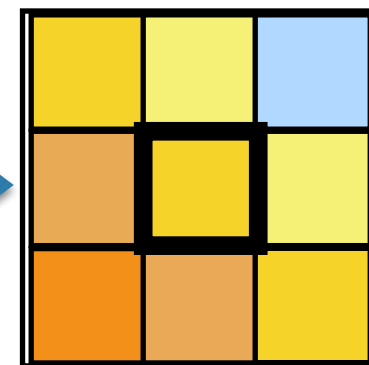
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



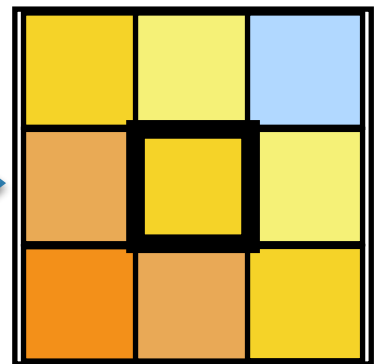
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

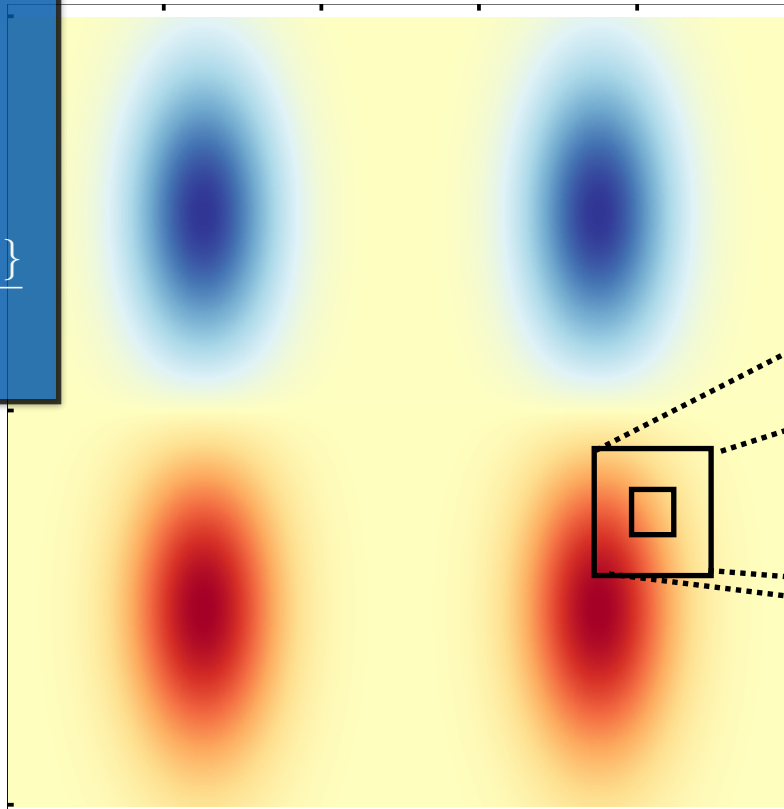
$w$

Small scales:

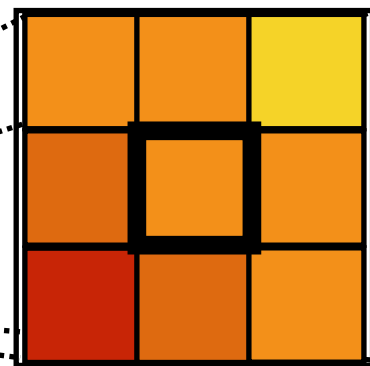
$\sigma \dot{B}$

Variance  
tensor:

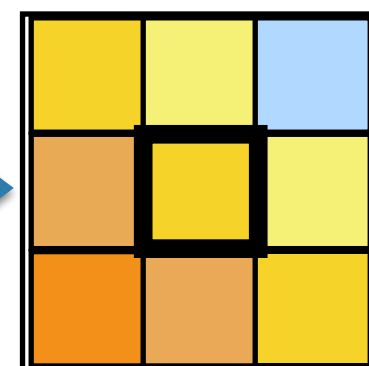
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



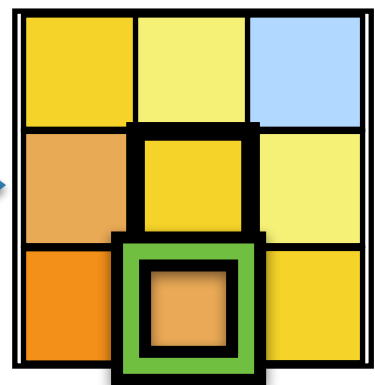
Neighbor



Centered  
neighbor



Random  
selection



# Random switching of points

(heterogeneous and non-stationary  $\sigma \dot{B}$ )

MU SVD

Large scales:

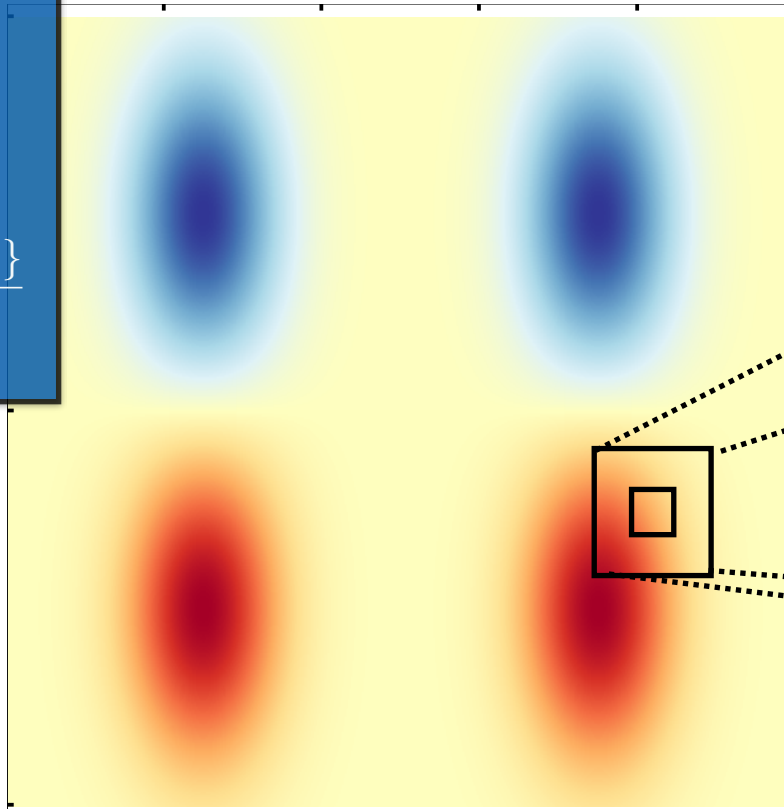
$w$

Small scales:

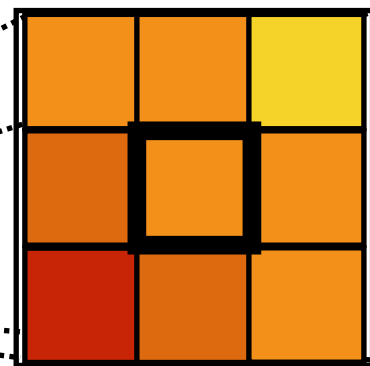
$\sigma \dot{B}$

Variance  
tensor:

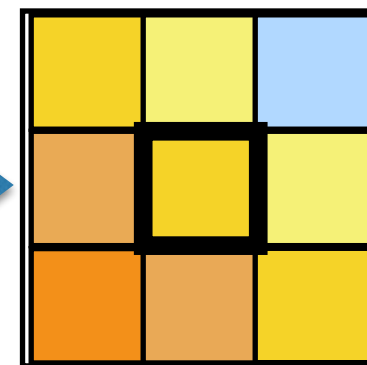
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



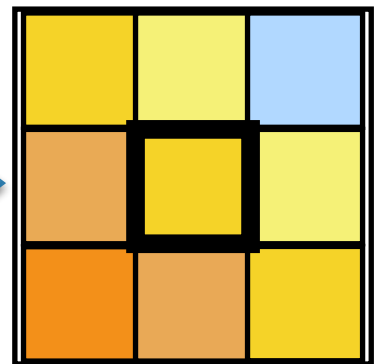
Neighbor



Centered  
neighbor



Random  
selection





# Random switching of points

MU SVD

Large scales:

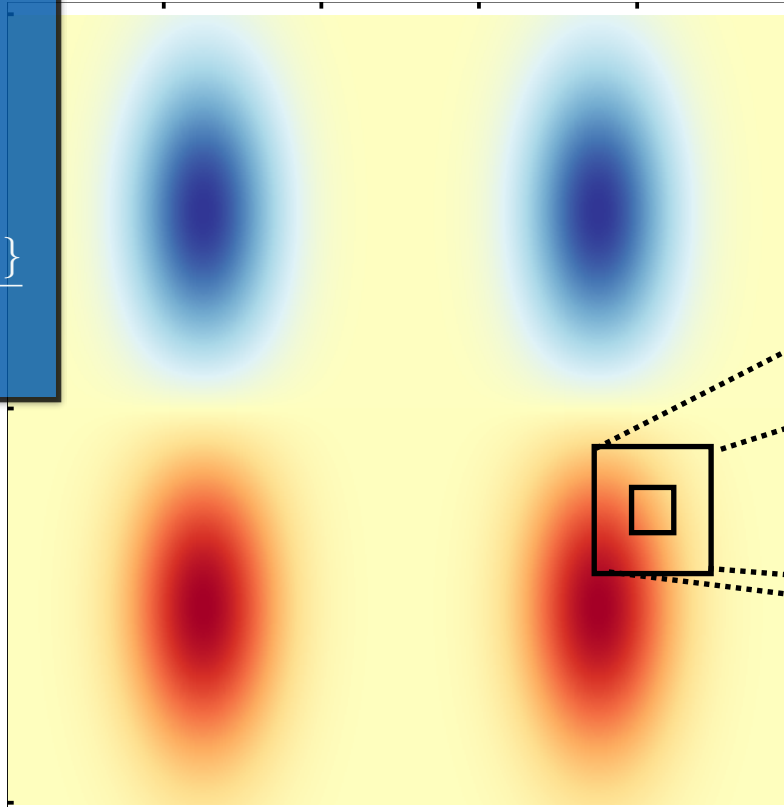
$w$

Small scales:

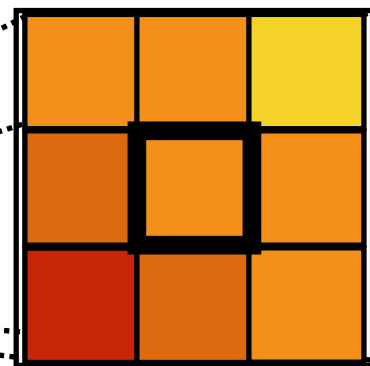
$\sigma \dot{B}$

Variance  
tensor:

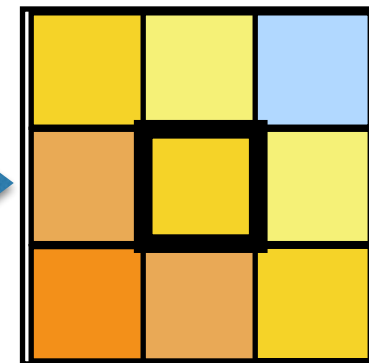
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



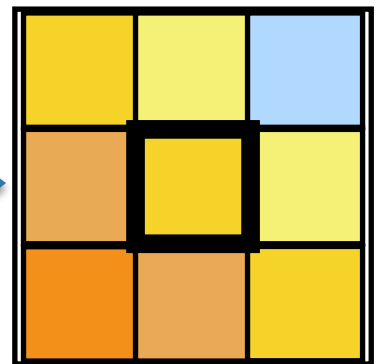
Neighbor



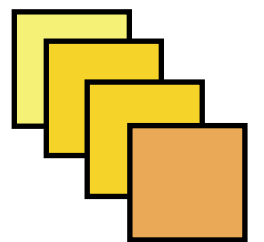
Centered  
neighbor



Random  
selection



Local ensemble :



# Random switching of points

MU SVD

Large scales:

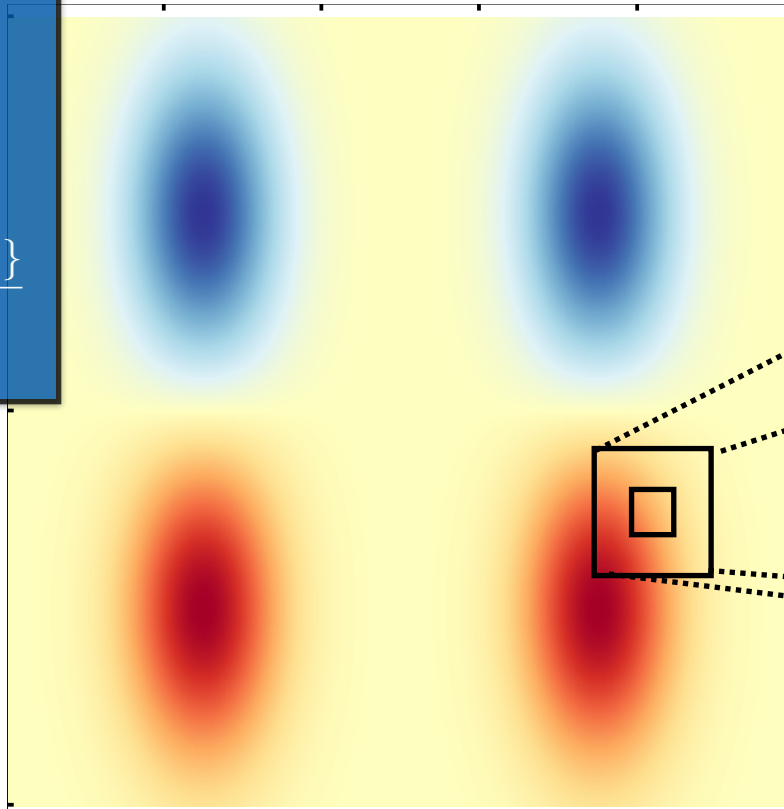
$w$

Small scales:

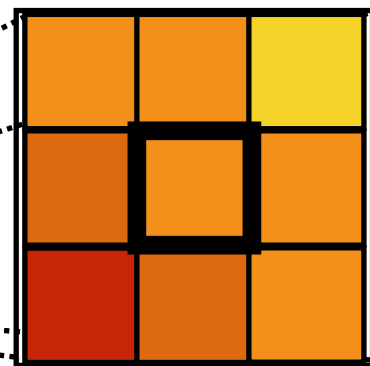
$\sigma \dot{B}$

Variance  
tensor:

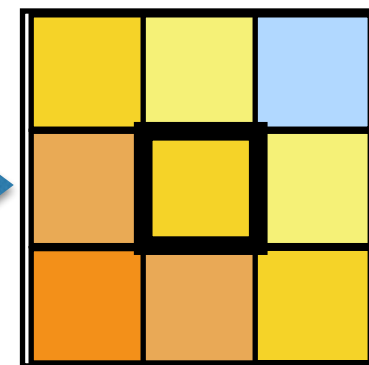
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



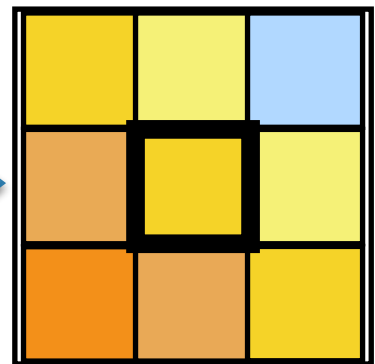
Neighbor



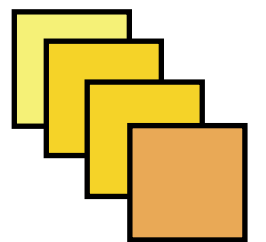
Centered  
neighbor



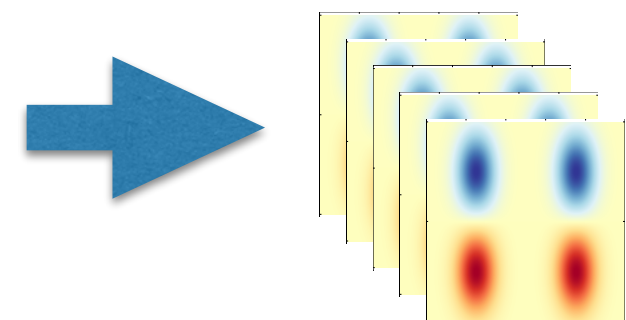
Random  
selection



Local ensemble :



Global ensemble



# Random switching of points

MU SVD

Large scales:

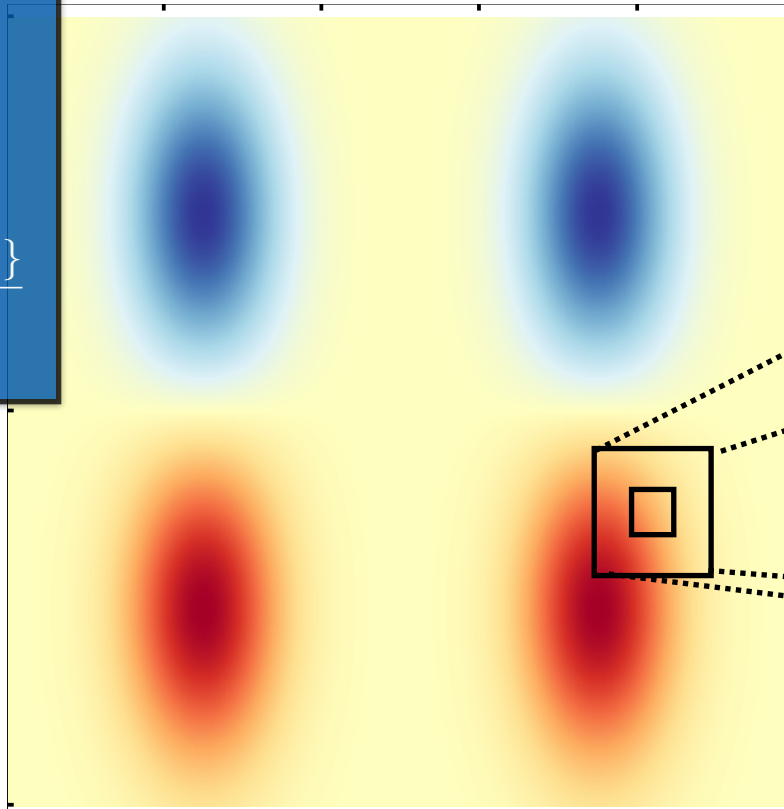
$w$

Small scales:

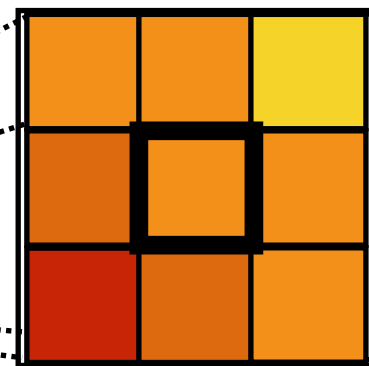
$\sigma \dot{B}$

Variance  
tensor:

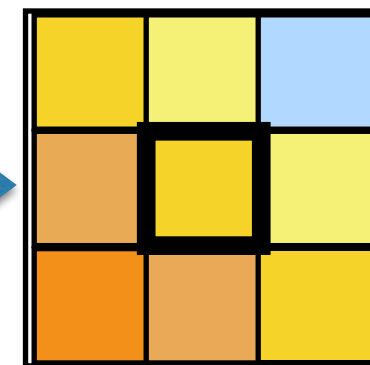
$$a = a(x, x) = \frac{\mathbb{E}\{\sigma dB (\sigma dB)^T\}}{dt}$$



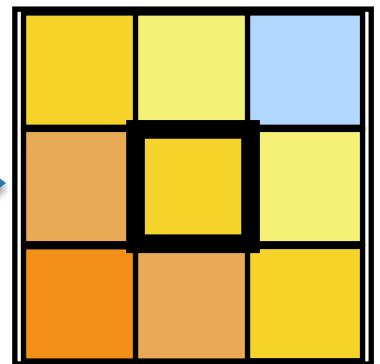
Neighbor



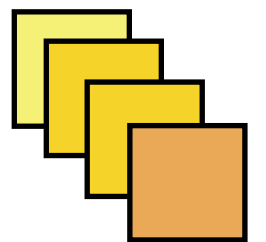
Centered  
neighbor



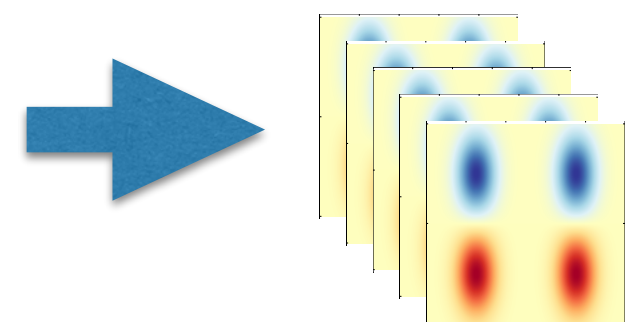
Random  
selection



Local ensemble :



Global ensemble



SVD



$\sigma \dot{B}$

# Part III

A new energy-budget-based  
stochastic scheme:

## WaveHyperv


# WaveHyperv

Transport  
equation

$$\frac{Dq}{Dt} = \mathcal{L}[q] + \eta$$

# WaveHyperv

Transport equation

$$\frac{Dq}{Dt} = \mathcal{L}[q] + \eta$$


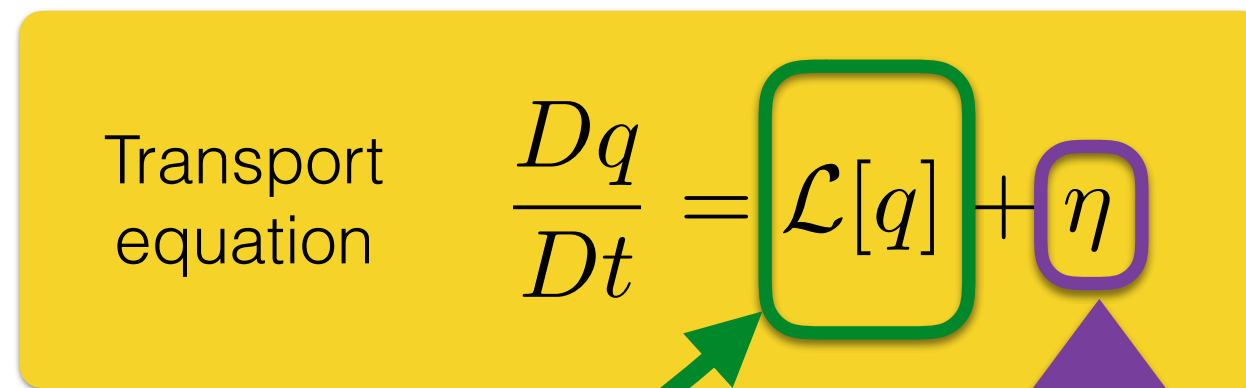
Usual  
deterministic  
subgrid tensor

e.g. Hyper-viscosity

$$\mathcal{L}[q] = -\nu \Delta^4 q$$

# WaveHyperv

Transport equation

$$\frac{Dq}{Dt} = \mathcal{L}[q] + \eta$$


Usual  
deterministic  
subgrid tensor

e.g. Hyper-viscosity

$$\mathcal{L}[q] = -\nu \Delta^4 q$$

Random forcing

built to meet the energy budget:

$$(\text{Random energy intake}) = \zeta \times \text{Dissipation}$$

# Summary of UQ methods

Name	Method
<b>MU Spec</b>	LU with homogeneous and stationary small-scale velocity
<b>MU ADSD</b>	LU with homogeneous, non-stationary and tuning-free small-scale velocity
<b>MU SVD</b>	LU with inhomogeneous and non-stationary small-scale velocity
<b>WaveHyperv</b>	Energy-budget-based stochastic scheme
<b>PIC Spec</b>	Perturbed initial conditions with homogeneous noise
<b>PIC SVD</b>	Perturbed initial conditions with inhomogeneous noise



# Part IV

## Numerical comparisons

# Test case 1:

$t = 17 \text{ days}$

**SQG**

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

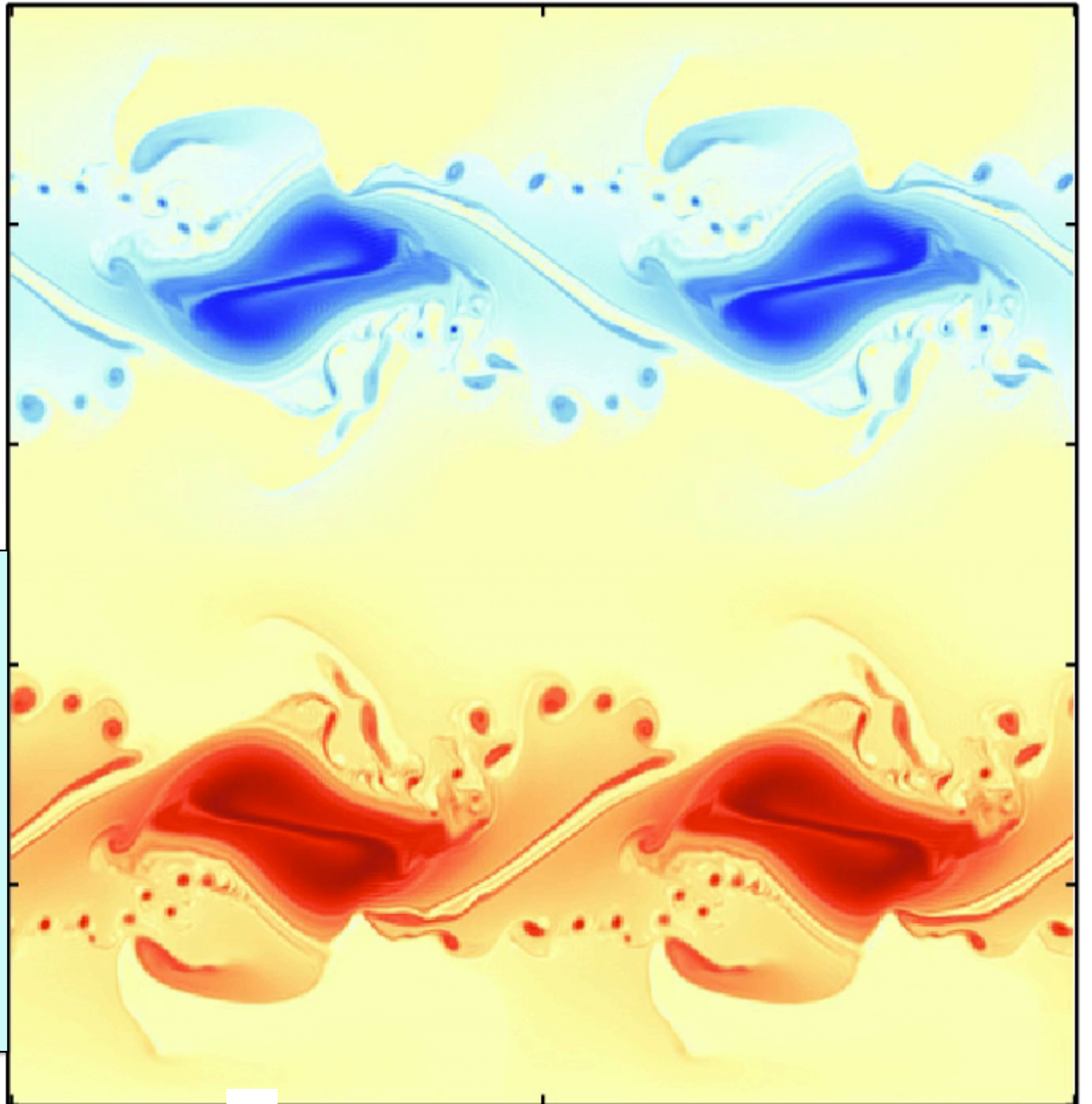
$$\mathbf{u} = \left( \text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic

SQG

512 x 512



# Test case 1:

$t = 17 \text{ days}$

**SQG**

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

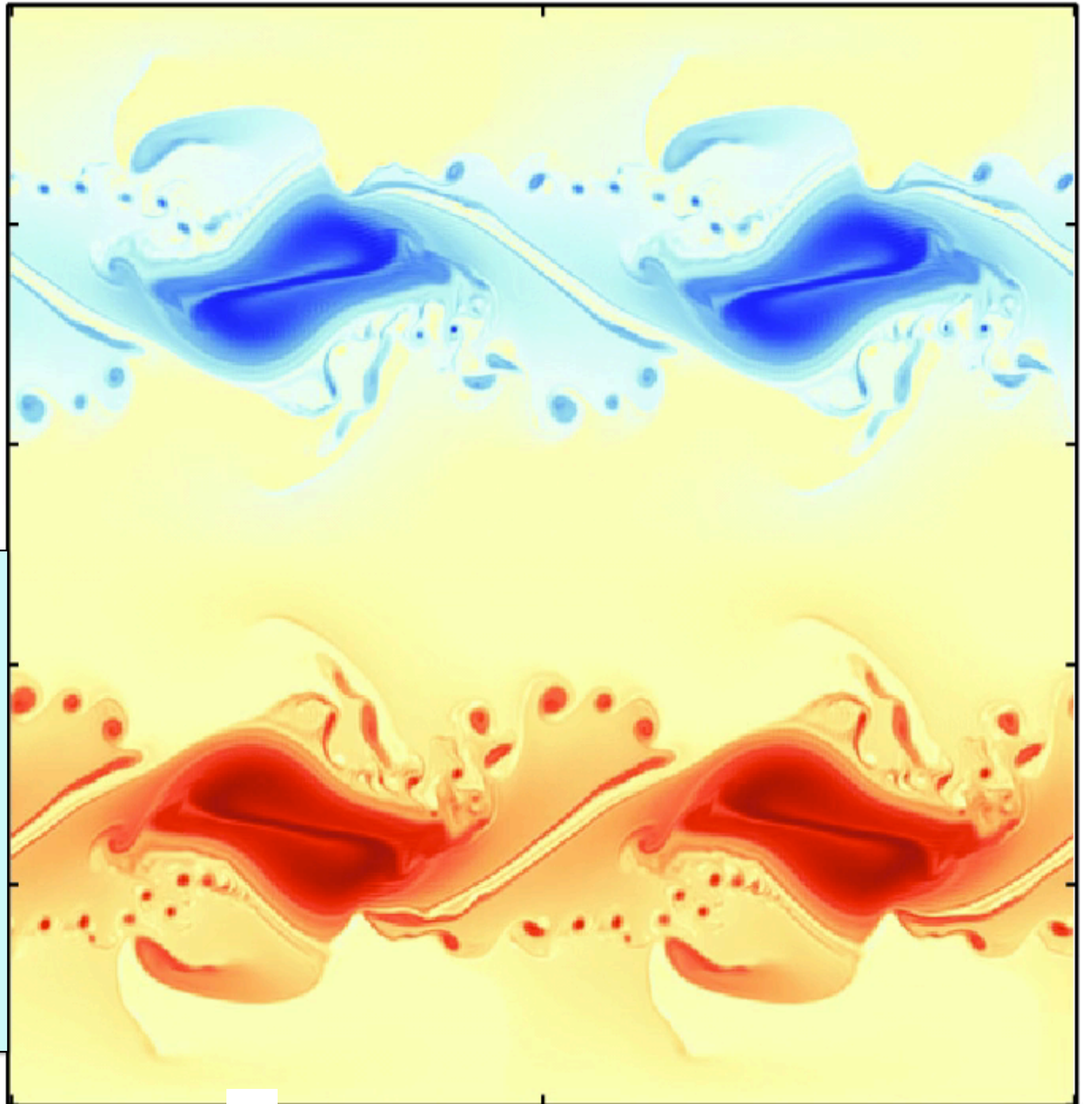
$$\mathbf{u} = \left( \text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic

SQG

512 x 512





# Test case 2:

$t = 10$  days

**SQG**

$$\frac{Db}{Dt} = -\alpha_{HV} \Delta^4 b \quad \text{Hyper-viscosity}$$

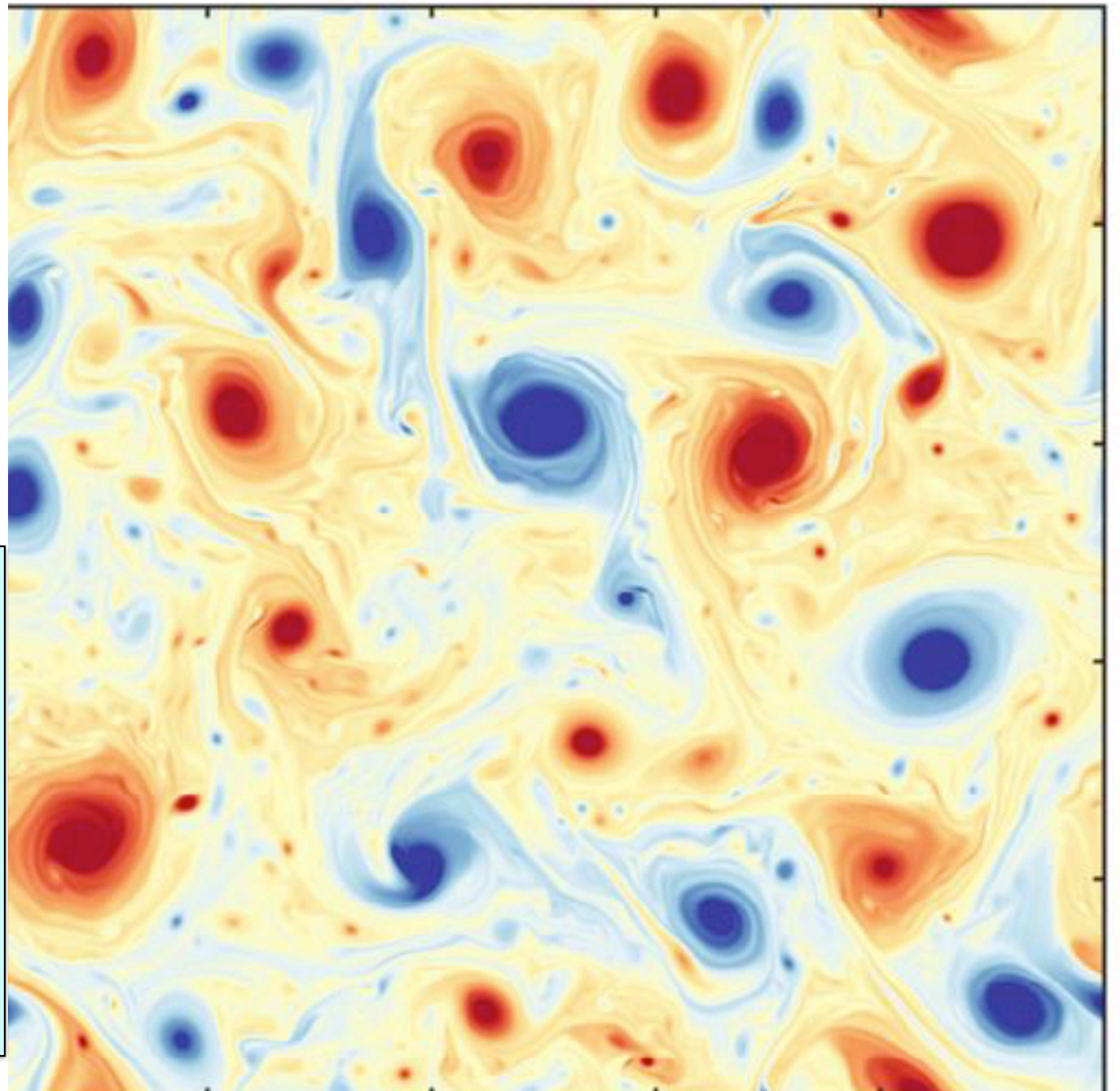
$$\mathbf{u} = \left( \text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic

SQG

512 x 512





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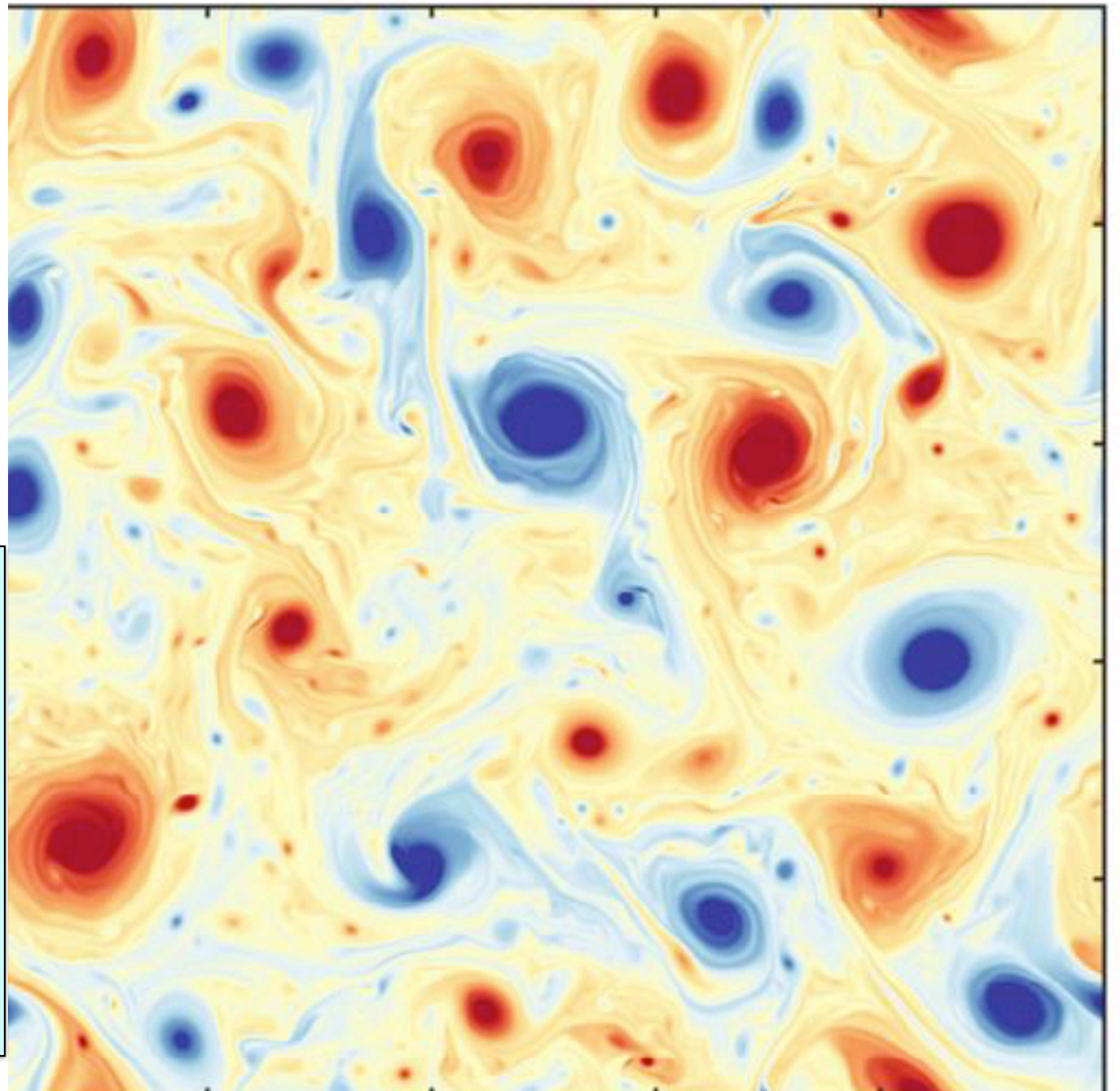
$$\mathbf{u} = \left( \text{cst.} \nabla^\perp \Delta^{-\frac{1}{2}} \right) b$$

Reference flow:

deterministic

SQG

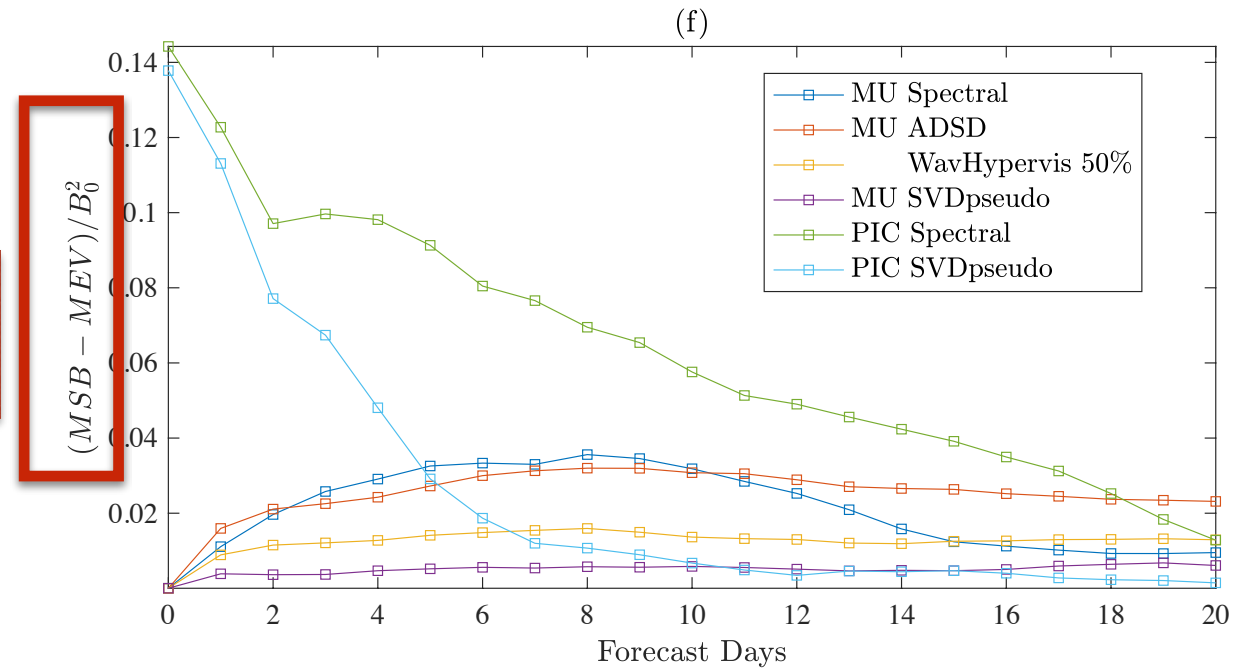
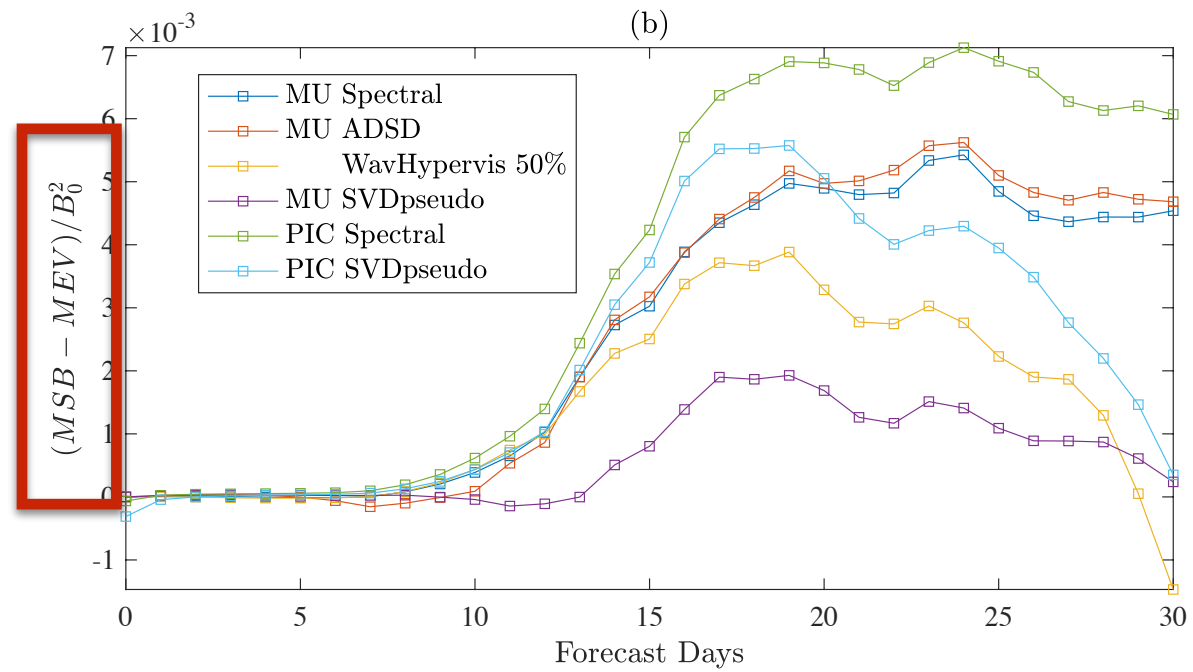
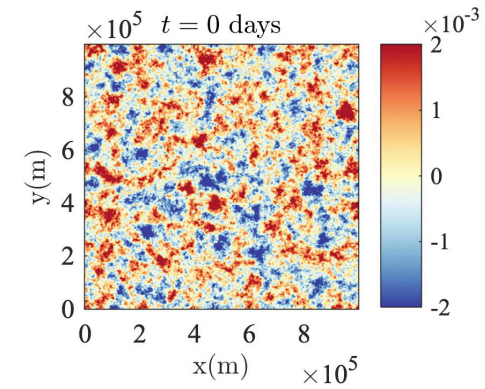
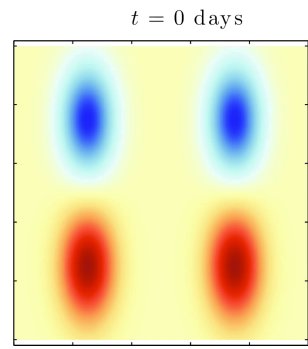
512 x 512



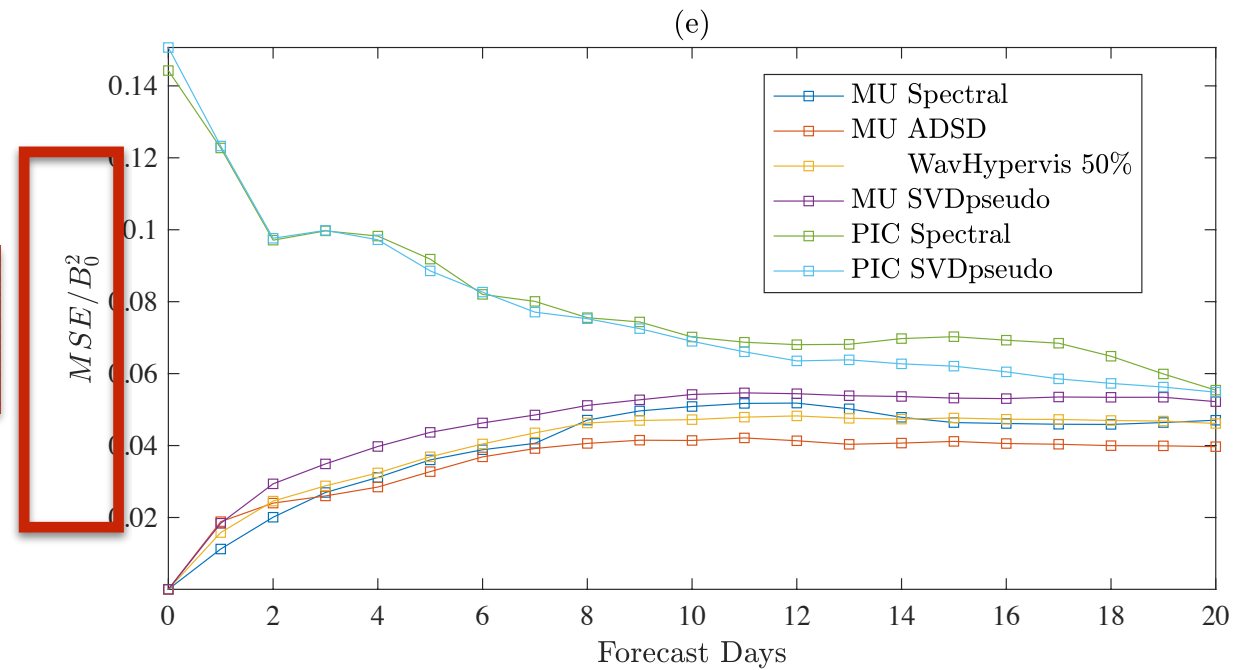
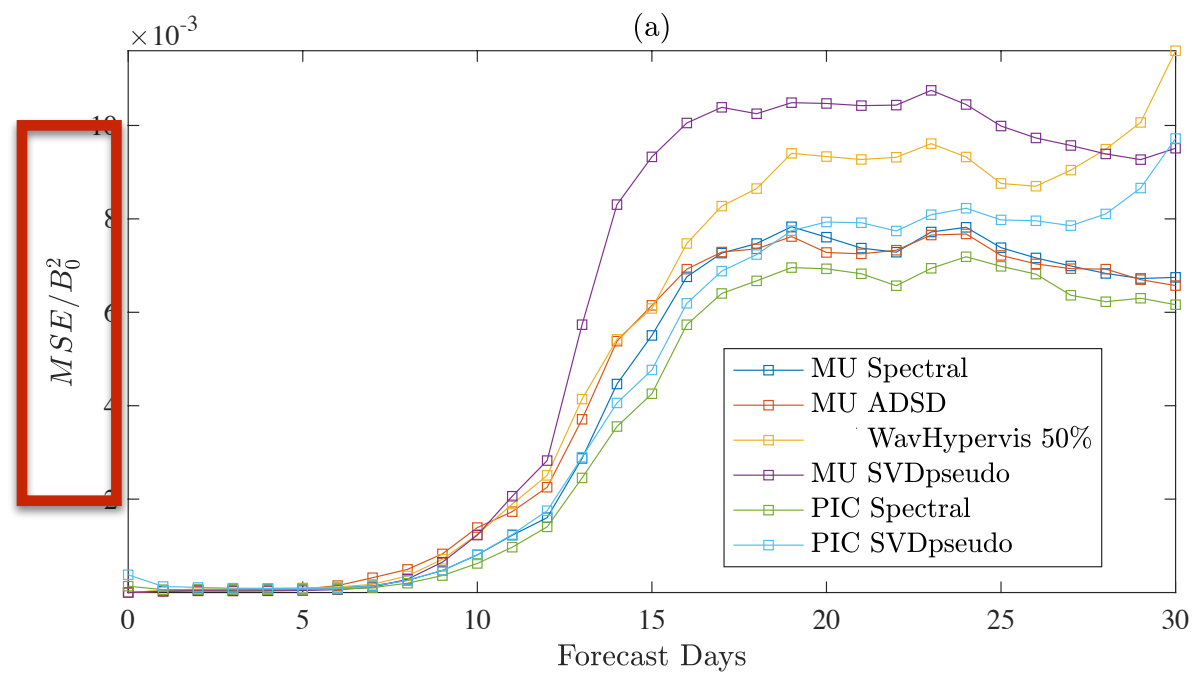
# UQ metrics

Metric	Meaning
RMSE	<u>Error</u> of ensemble members
Talagrand histogram (TH)	Capacity of the ensemble to <u>explore</u> all reference possible values
Bias <sup>2</sup> -spread	Capacity of the ensemble to <u>explore</u> all reference possible values
CRPS	Point-wise distance between the ensemble CFD and the indicator function of the event
Energy Score (ES)	Generalized CRPS for multivariate ensemble

# Spreading VS Errors



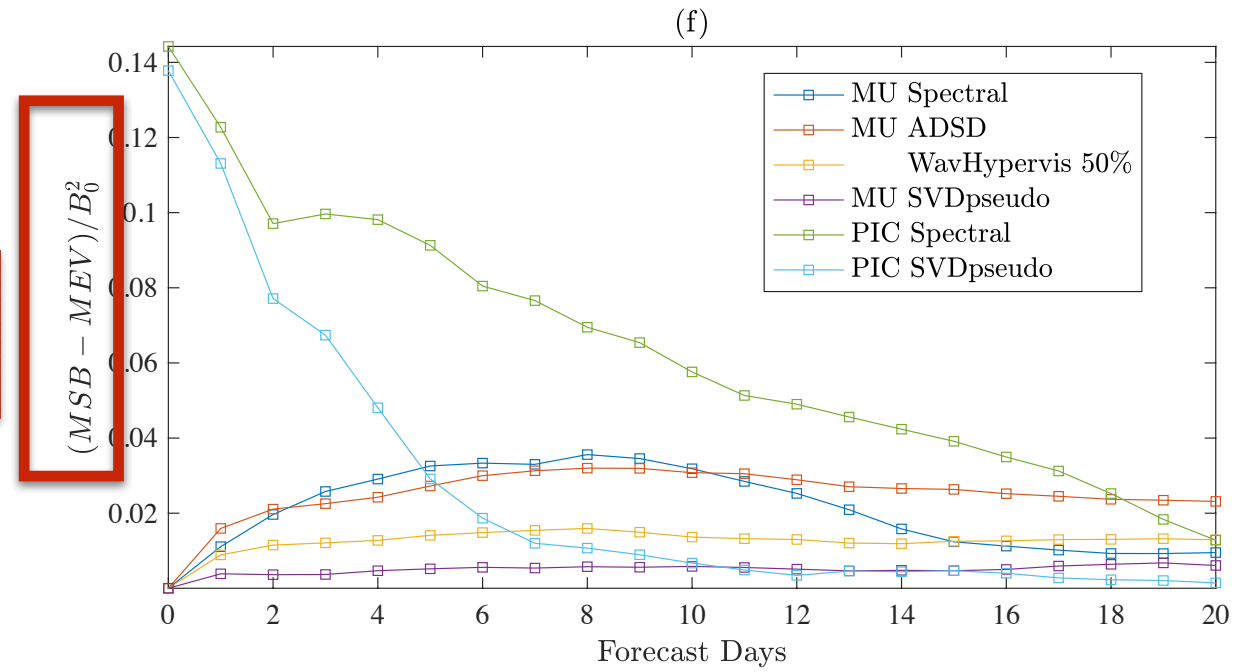
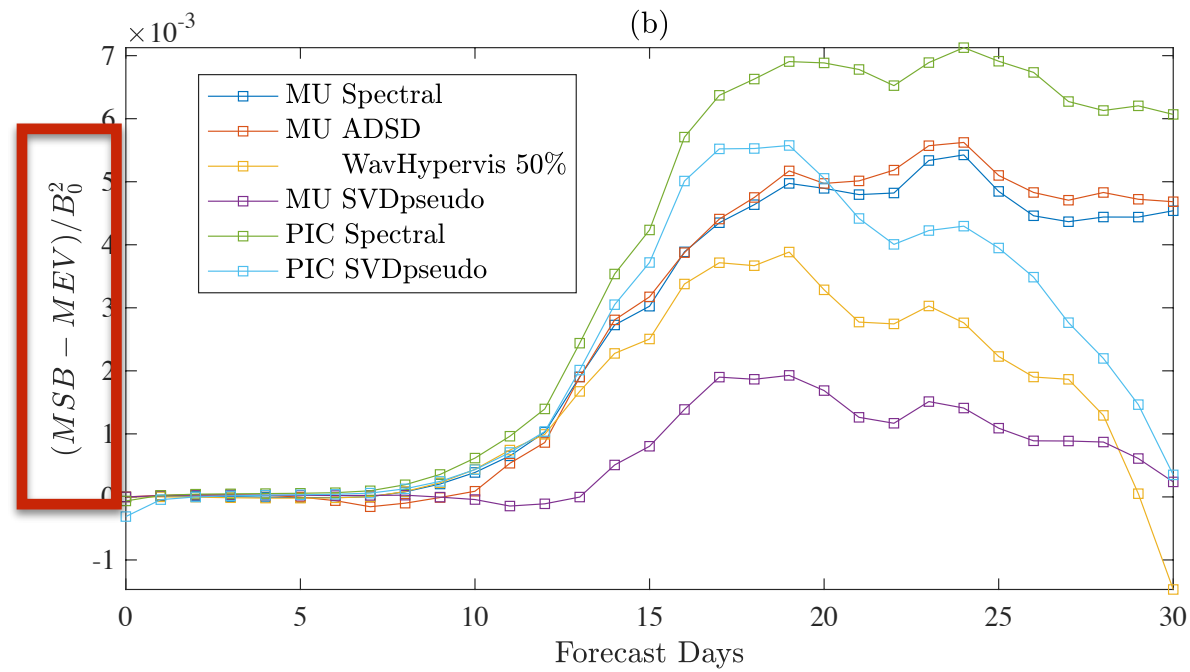
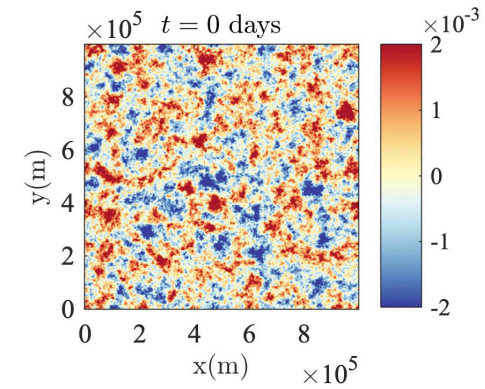
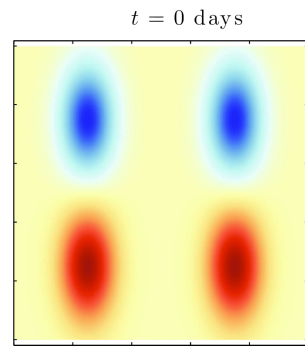
Spread



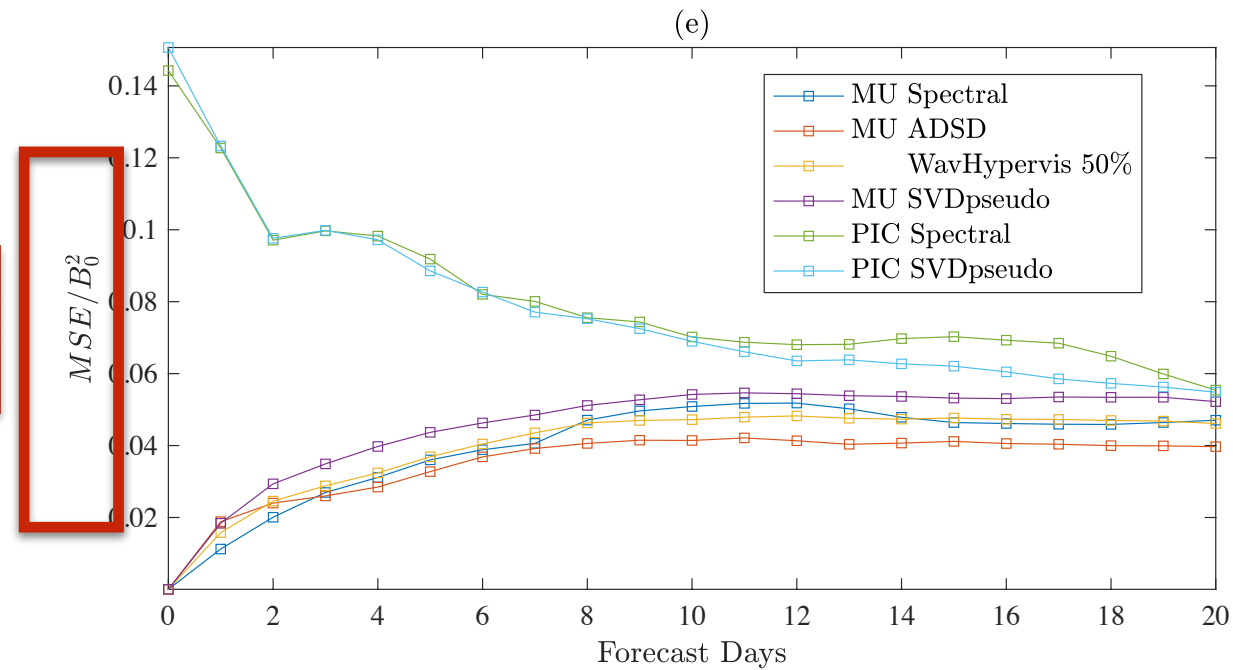
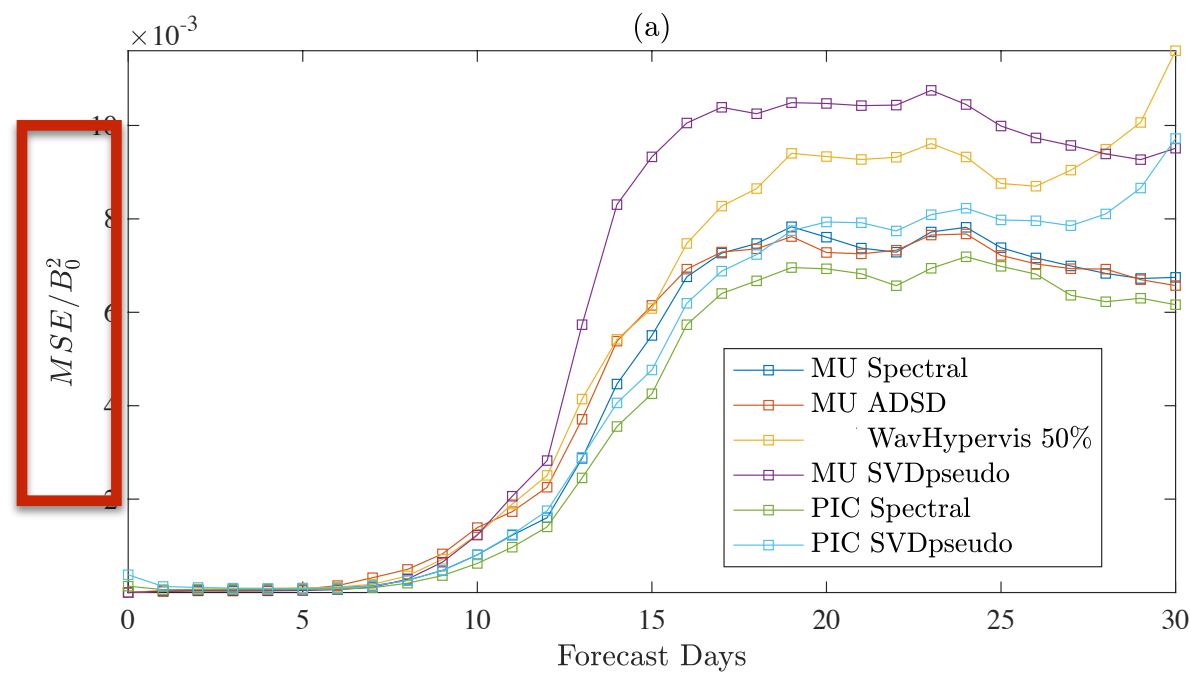
Error



# Spreading VS Errors



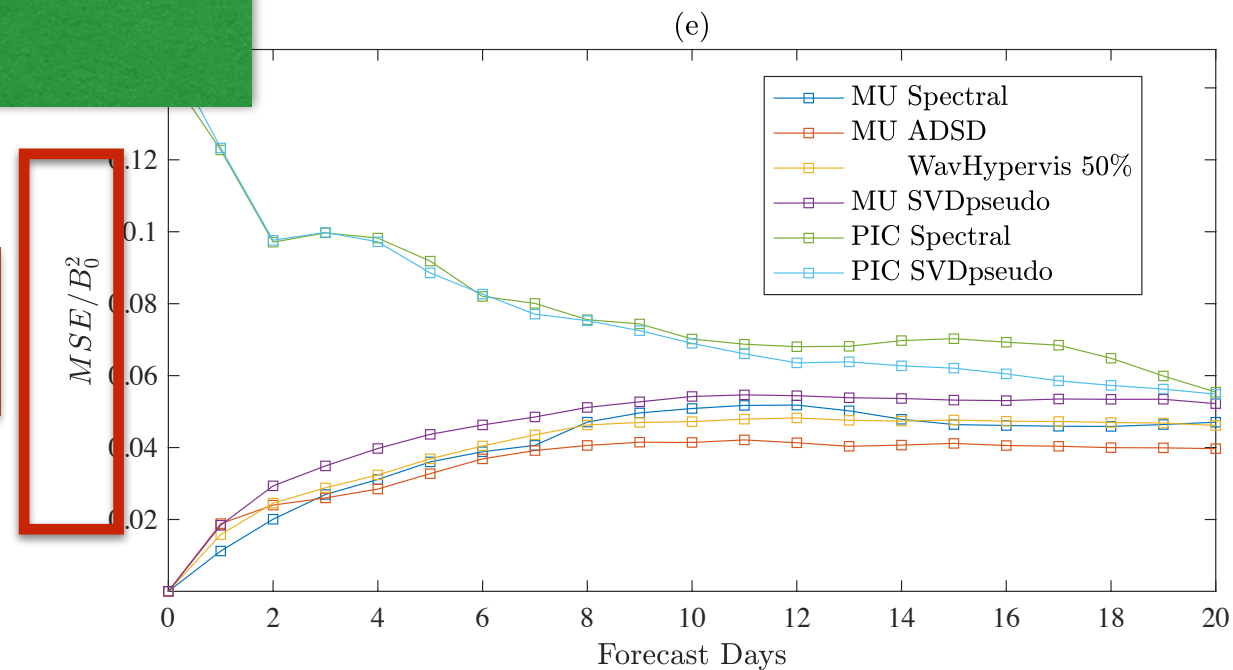
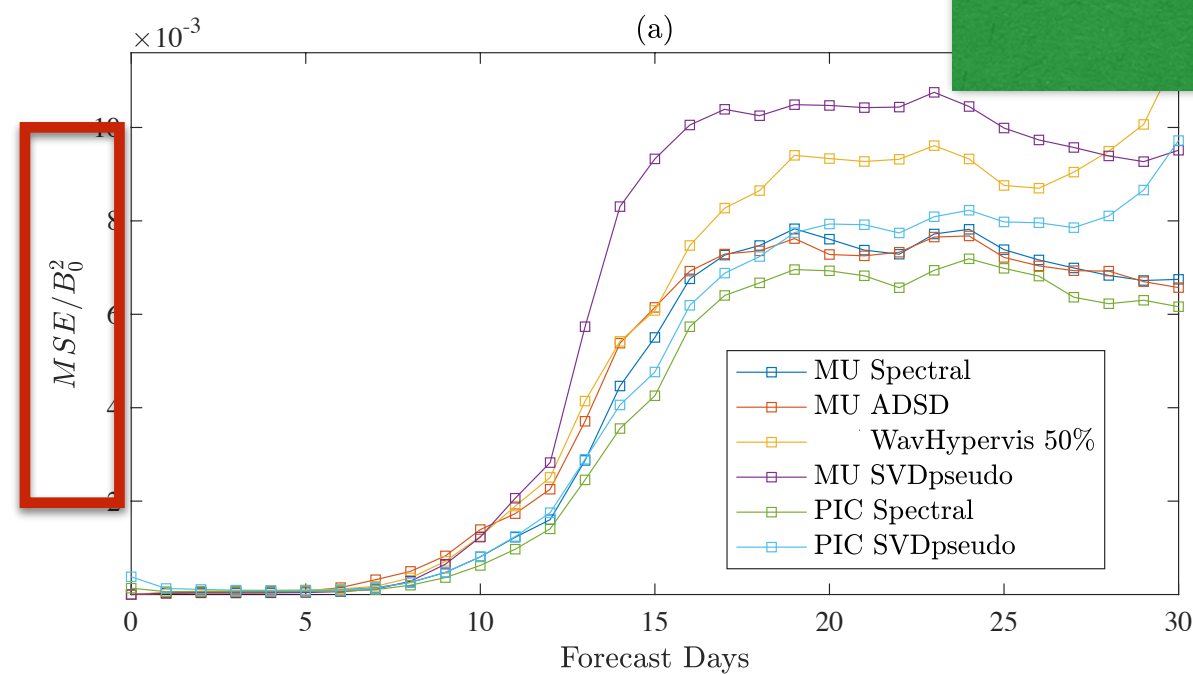
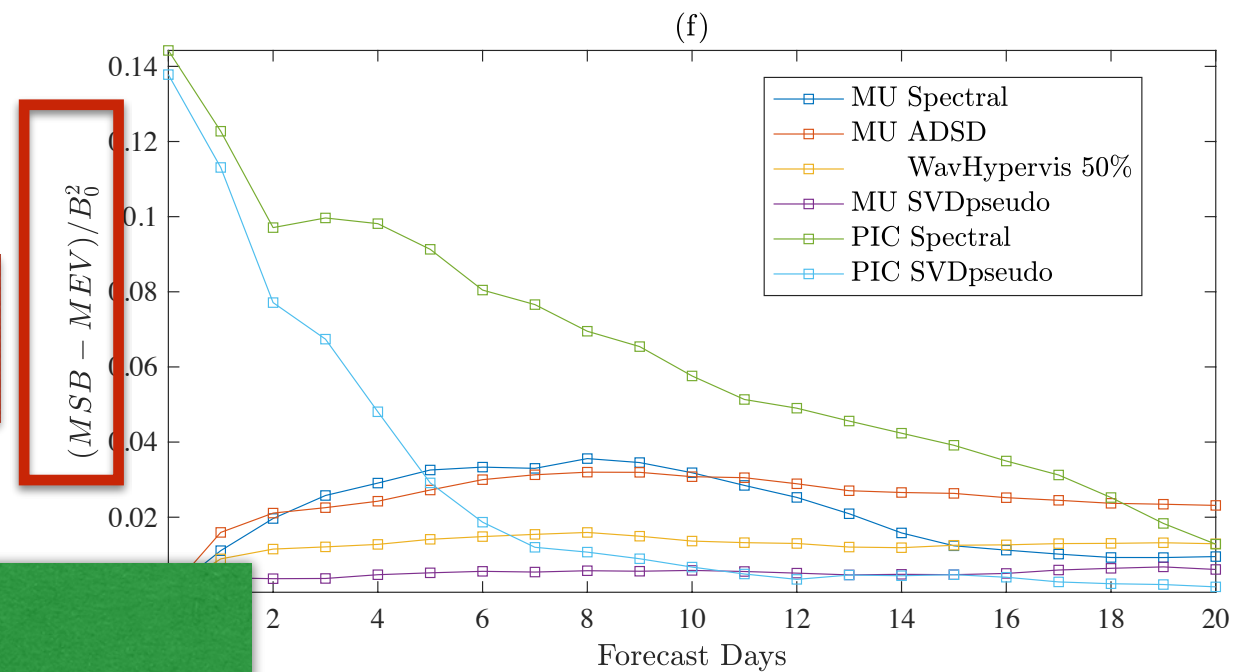
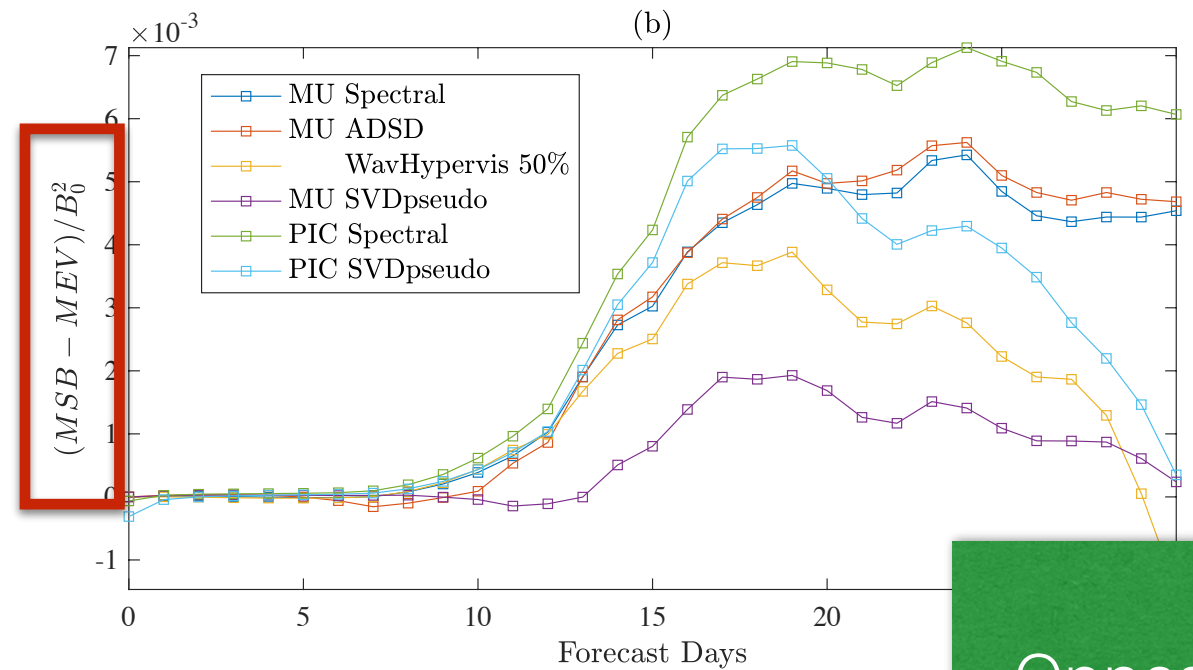
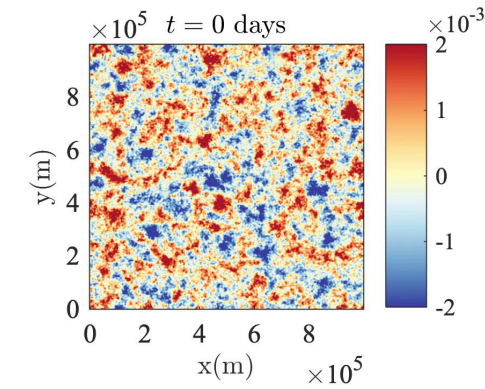
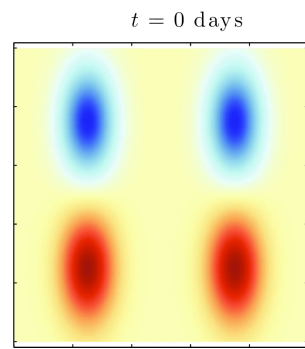
Spread



Error



# Spreading VS Errors



Spread

Opposite conclusions

Error

# Ensemble point-wise skills : CRPS

MU Spec

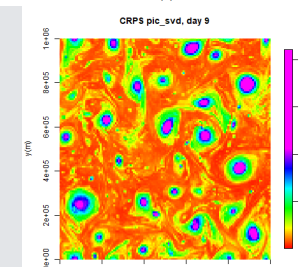
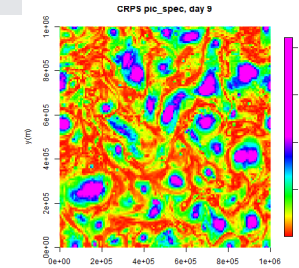
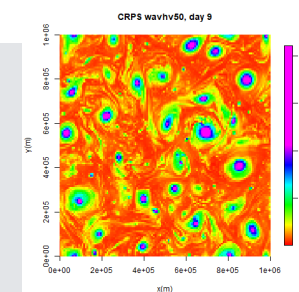
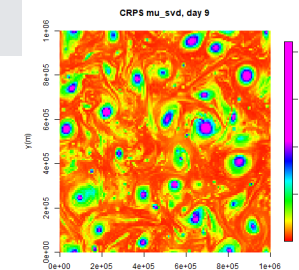
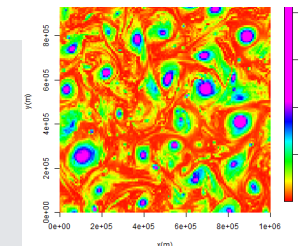
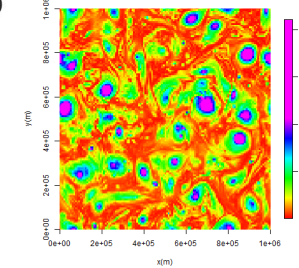
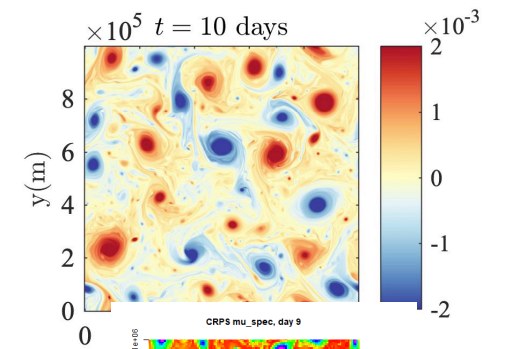
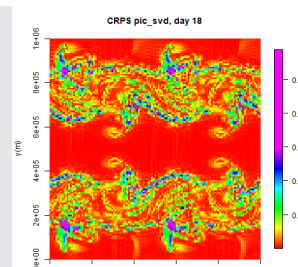
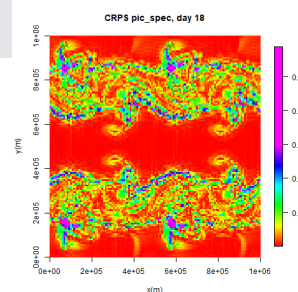
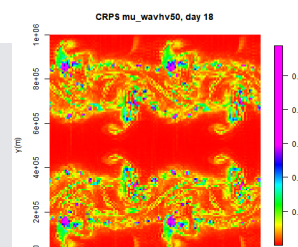
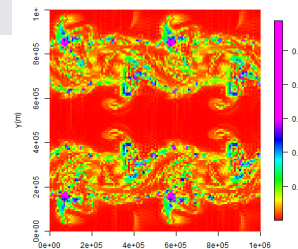
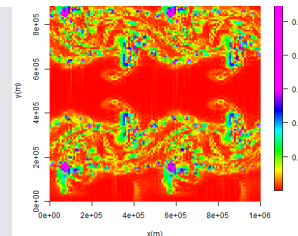
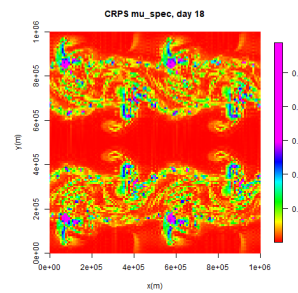
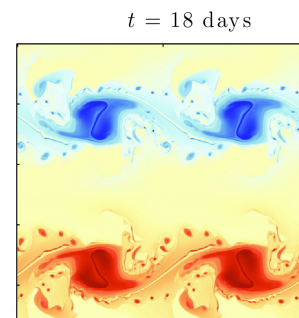
MU ADSD

MU SVD

WaveHyperv

PIC Spec

PIC SVD



# Ensemble point-wise skills : CRPS

MU Spec

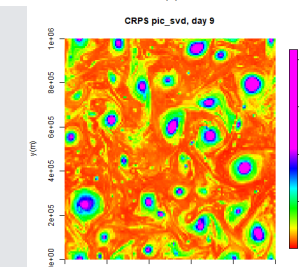
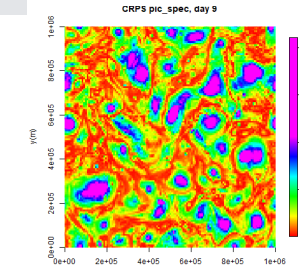
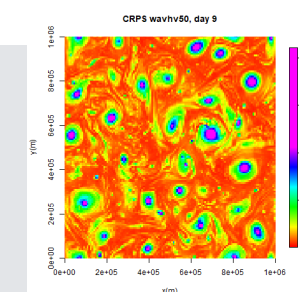
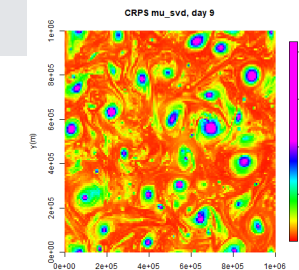
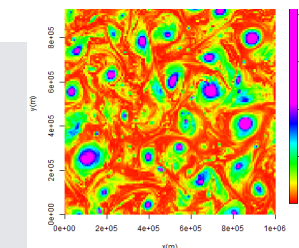
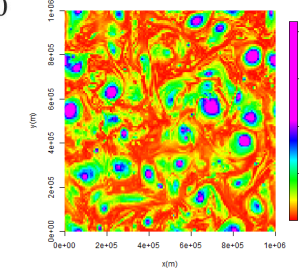
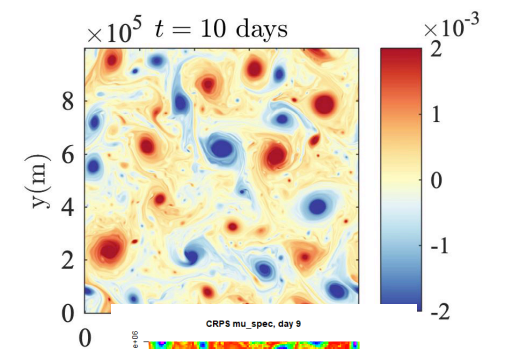
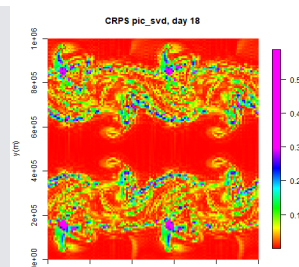
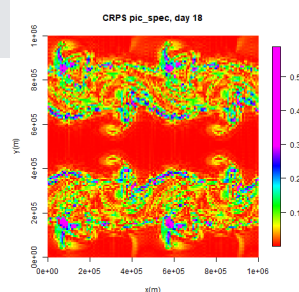
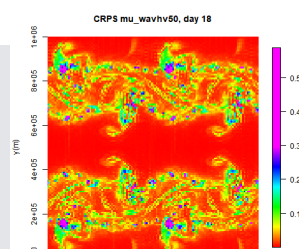
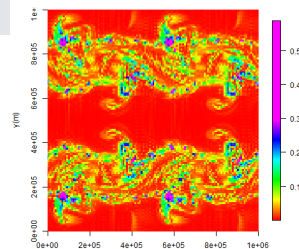
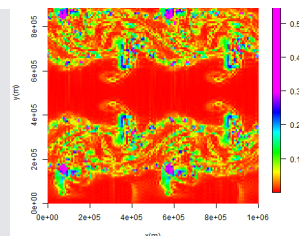
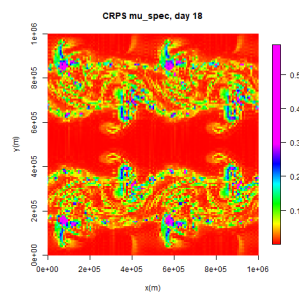
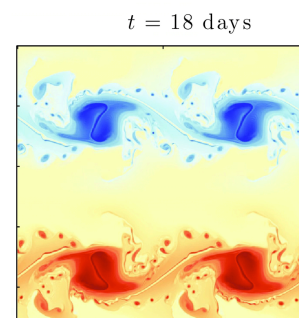
MU ADSD

MU SVD

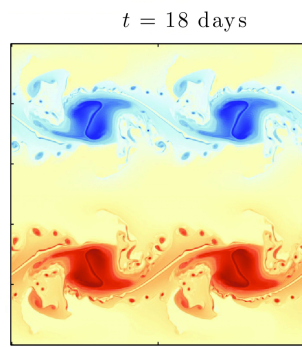
WaveHyperv

PIC Spec

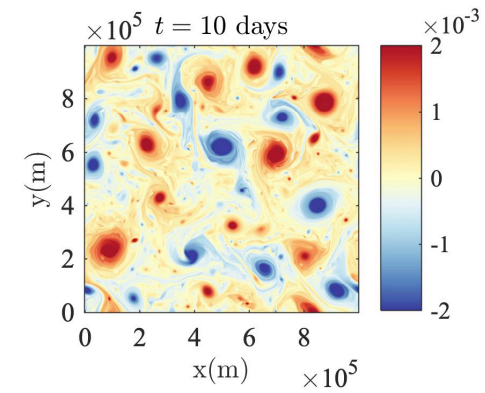
PIC SVD



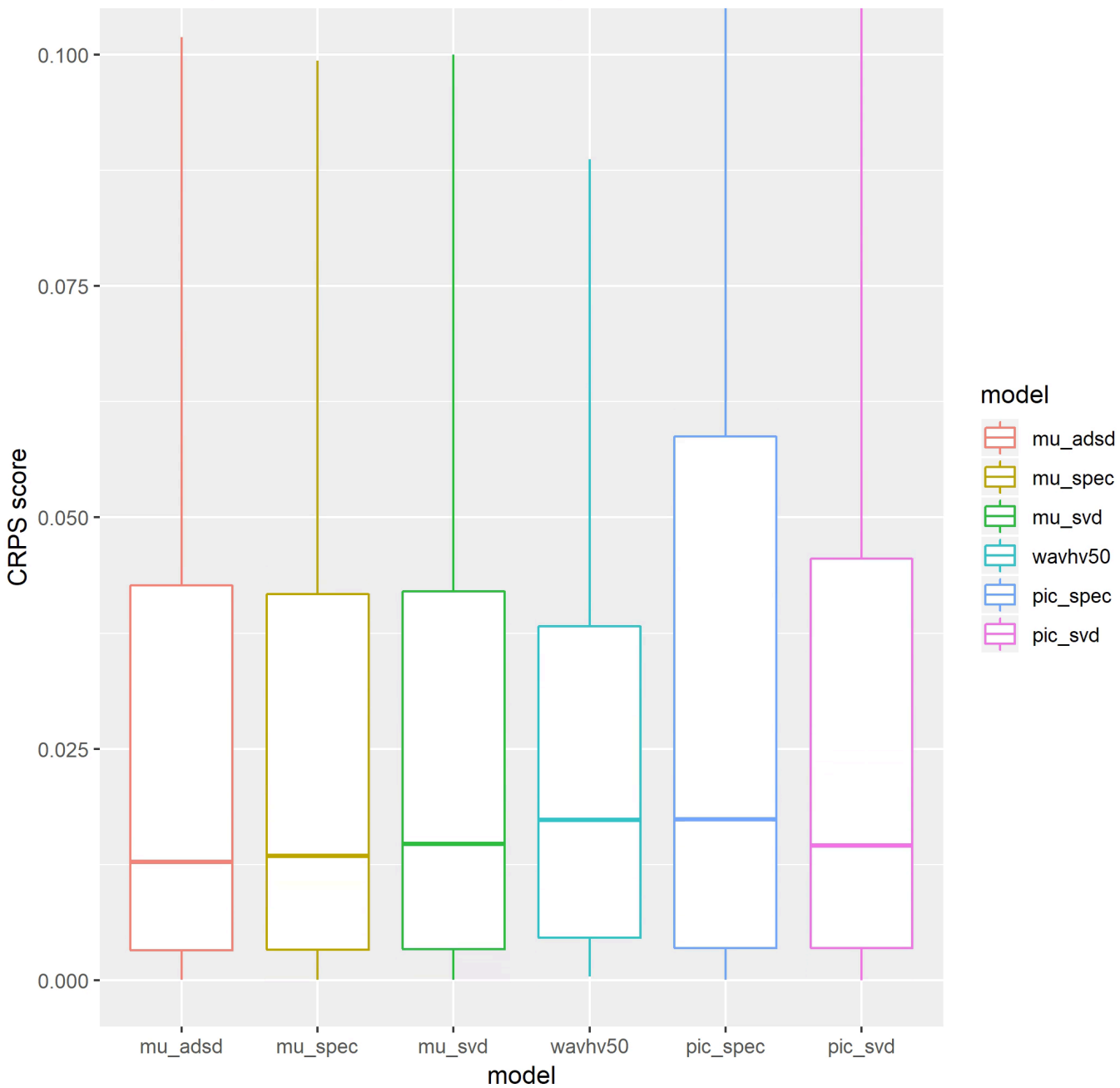




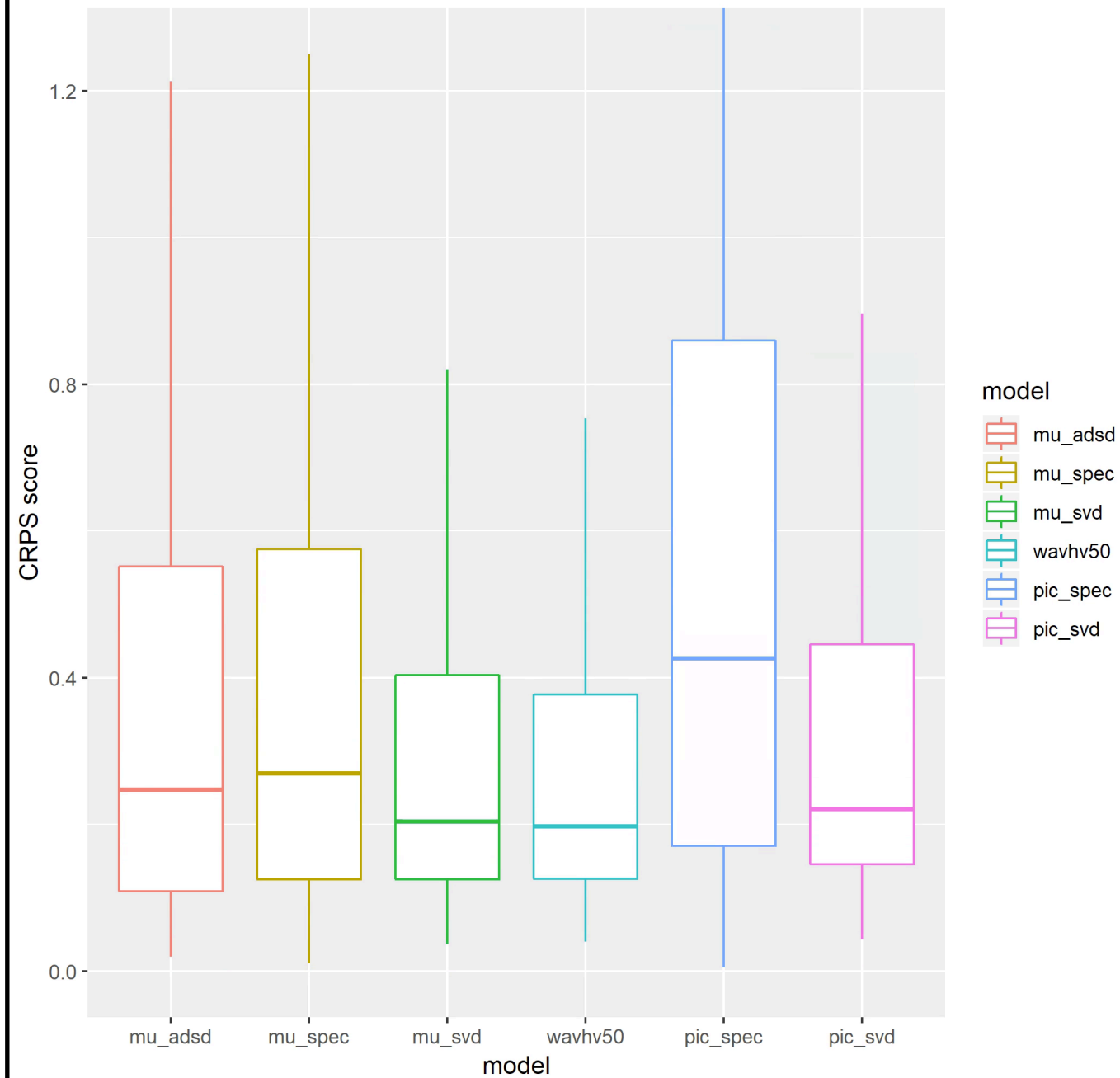
# Ensemble point-wise skills : CRPS

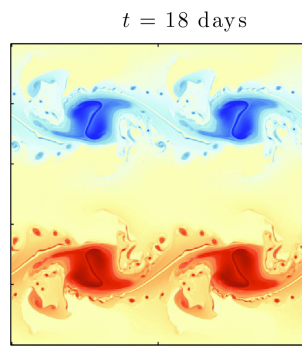


CRPS spatial of each model at day 19

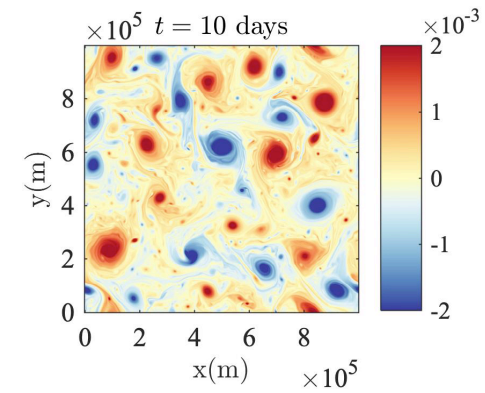


CRPS spatial of each model at day 10

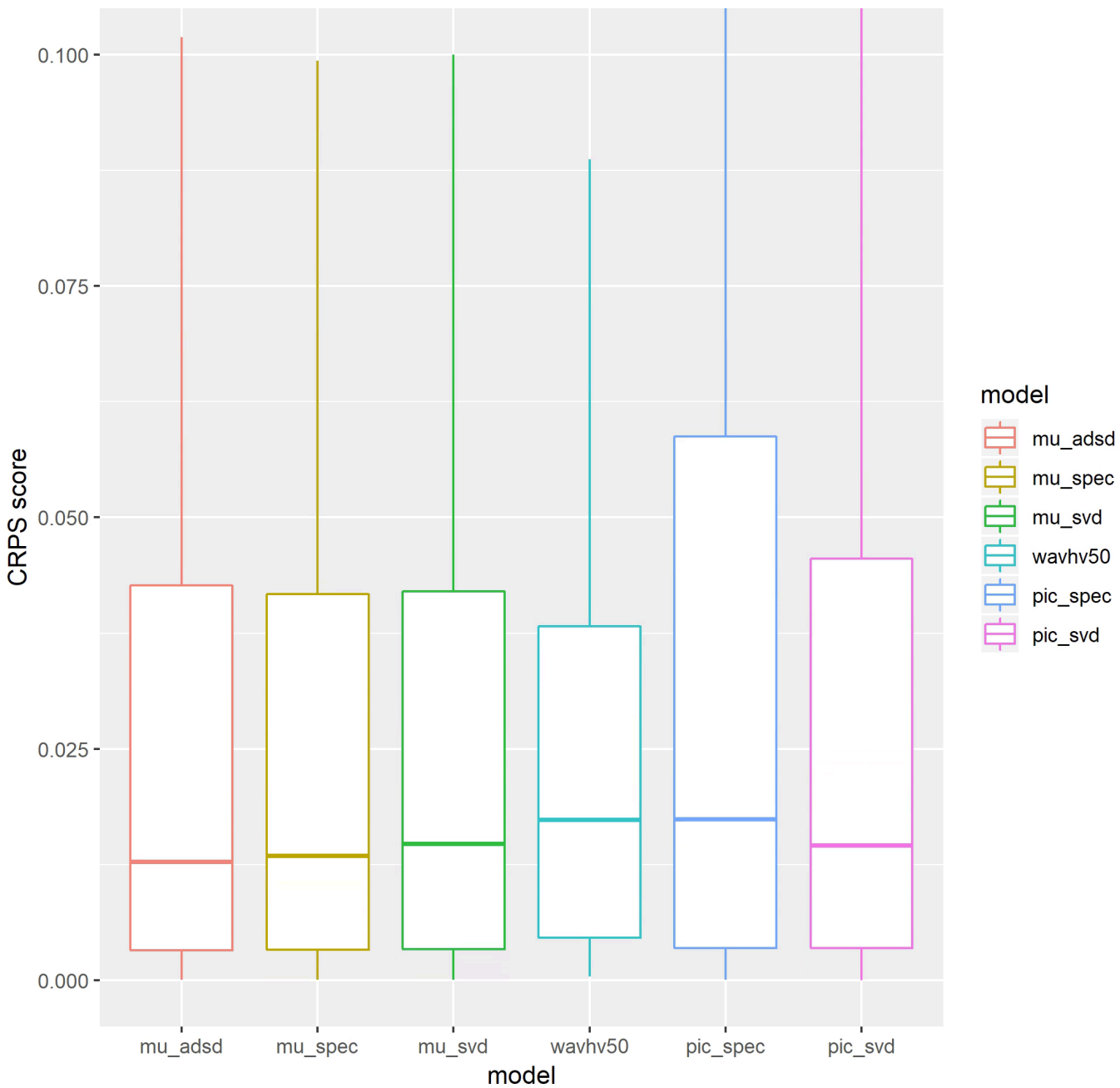




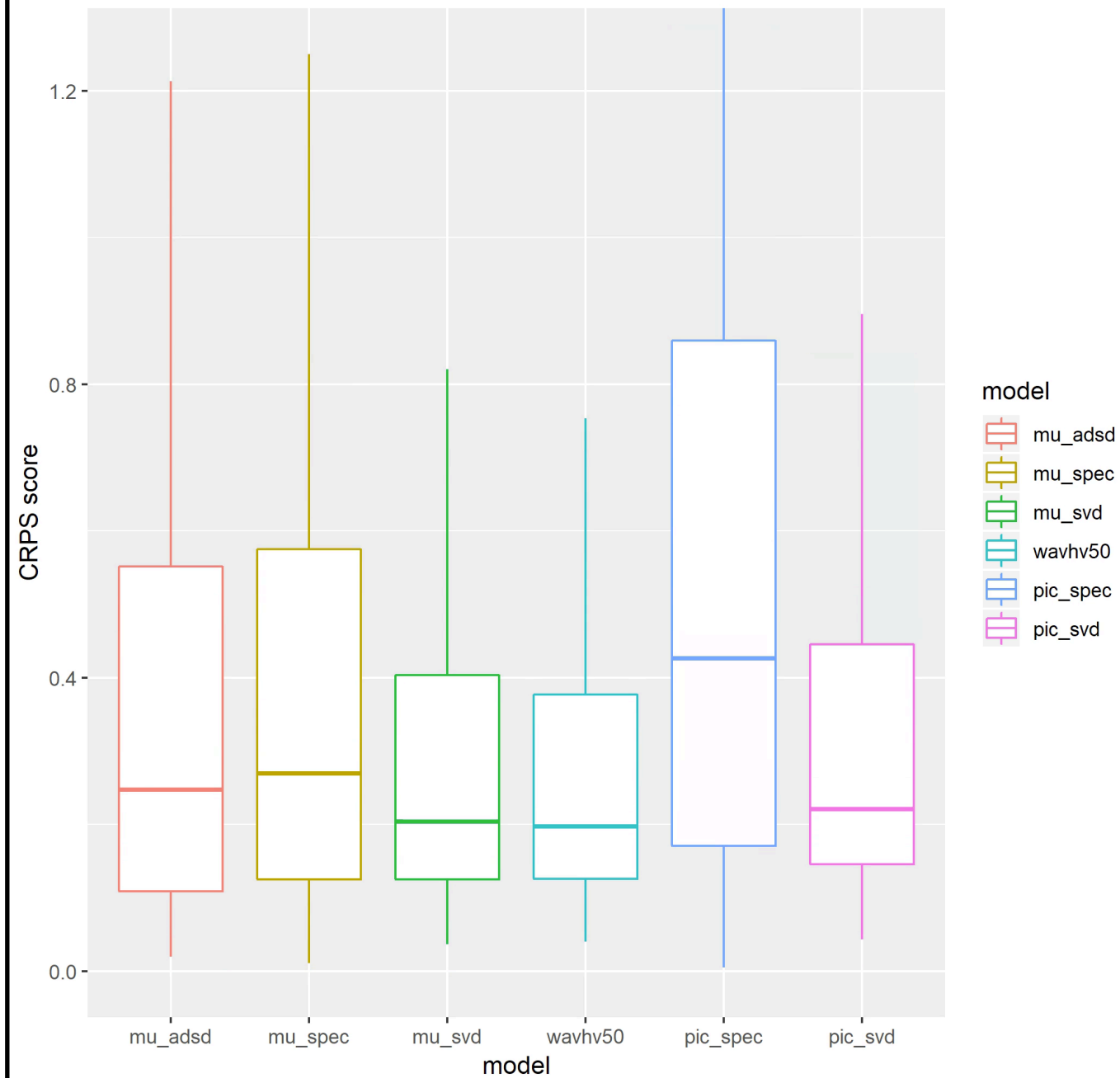
# Ensemble point-wise skills : CRPS



CRPS spatial of each model at day 19



CRPS spatial of each model at day 10



# Conclusion

# Conclusion

	RMSE (errors)	B <sup>2</sup> -Var (spread)	CRPS (point-wise)	ES (global)
MU Spec	+	+	+	+
MU ADSD	++	+	+	+
MU SVD	-	++	++	++
WaveHyperv	-	+	++	++
PIC Spec	-	--	--	--
PIC SVD	--	-	+	+

# Conclusion

	RMSE (errors)	B <sup>2</sup> -Var (spread)	CRPS (point-wise)	ES (global)
MU Spec	+	+	+	+
MU ADSD	++	+	+	+
MU SVD	-	++	++	++
WaveHyperv	-	+	++	++
<del>PIC Spec</del>	<del>-</del>	<del>--</del>	<del>--</del>	<del>--</del>
<del>PIC SVD</del>	<del>--</del>	<del>-</del>	<del>+</del>	<del>+</del>



# Conclusion

	RMSE (errors)	B <sup>2</sup> -Var (spread)	CRPS (point-wise)	ES (global)
MU Spec	+	+	+	+
MU ADSD	++	+	+	+
MU SVD	-	++	++	++
WaveHyperv	-	+	++	++
PIC Spec	-	--	--	--
PIC SVD	--	-	+	+

# Conclusion

	RMSE (errors)	B <sup>2</sup> -Var (spread)	CRPS (point-wise)	ES (global)
MU Spec	+	+	+	+
MU ADSD	++	+	+	+
MU SVD	-	++	++	++
WaveHyperv	-	+	++	++
PIC Spec	-	--	--	--
PIC SVD	--	-	+	+